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## Automatic Matching of Scans from Hamby Sets

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## Automatic Matching of Scans from Hamby Sets

### Disciplines

Forensic Science and Technology

### Comments

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# Automatic Matching of Scans from Hamby Sets Friends, False Friends and Clones

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Dr. Susan VanderPlas

Dr. Alicia Carriquiry

# An Investigation ... - the statistical way

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Dr. Susan VanderPlas

Dr. Alicia Carriquiry



# Outline

- ★ Matching bullets - very brief overview of the algorithm: *getting RF scores*
- ★ Applications for quantifying identifications: *using RF scores*
- ★ Scores as a diagnostics tool: *mis(?)using RF scores*

# Hamby Sets

- ★ in collaboration with: St Louis PD, DCI Ankeny, Jim Hamby, Paul Murphy, NIST NBTRD
- ★ Hamby Sets 10, 36, 44, 173, 224, 252
- ★ Hamby Clones 159
- ★ Hamby Clone Test Set 224
- ★ each set consists of
  - ★ 20 known bullets (2 from each of ten consecutively manufactured P-85 barrels)
  - ★ 15 questioned bullets

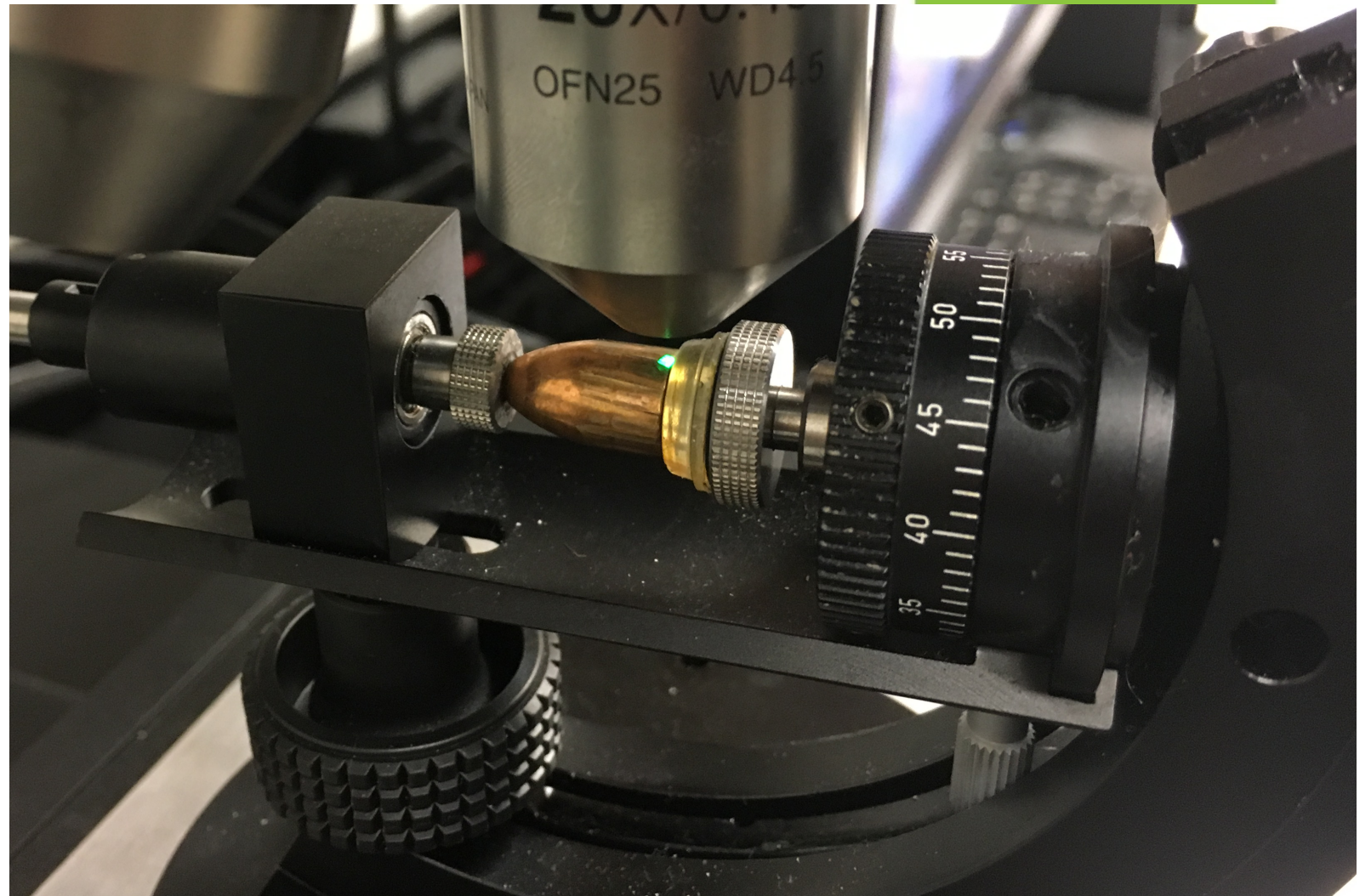
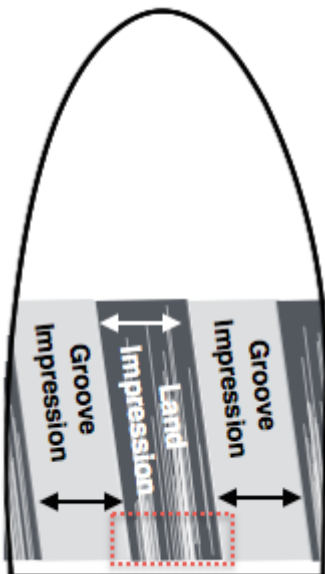
# Data: Microscope Facility

Roy J. Carver High Resolution Microscopy Facility

Two Sensofar Confocal  
Light Microscopes

Six undergraduates  
scanning bullet lands

3d topographic images:  
height measurements on  
x-y grid





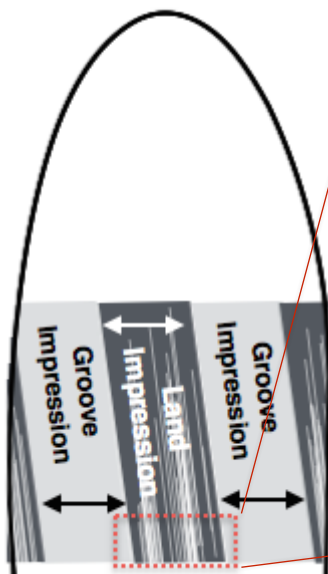
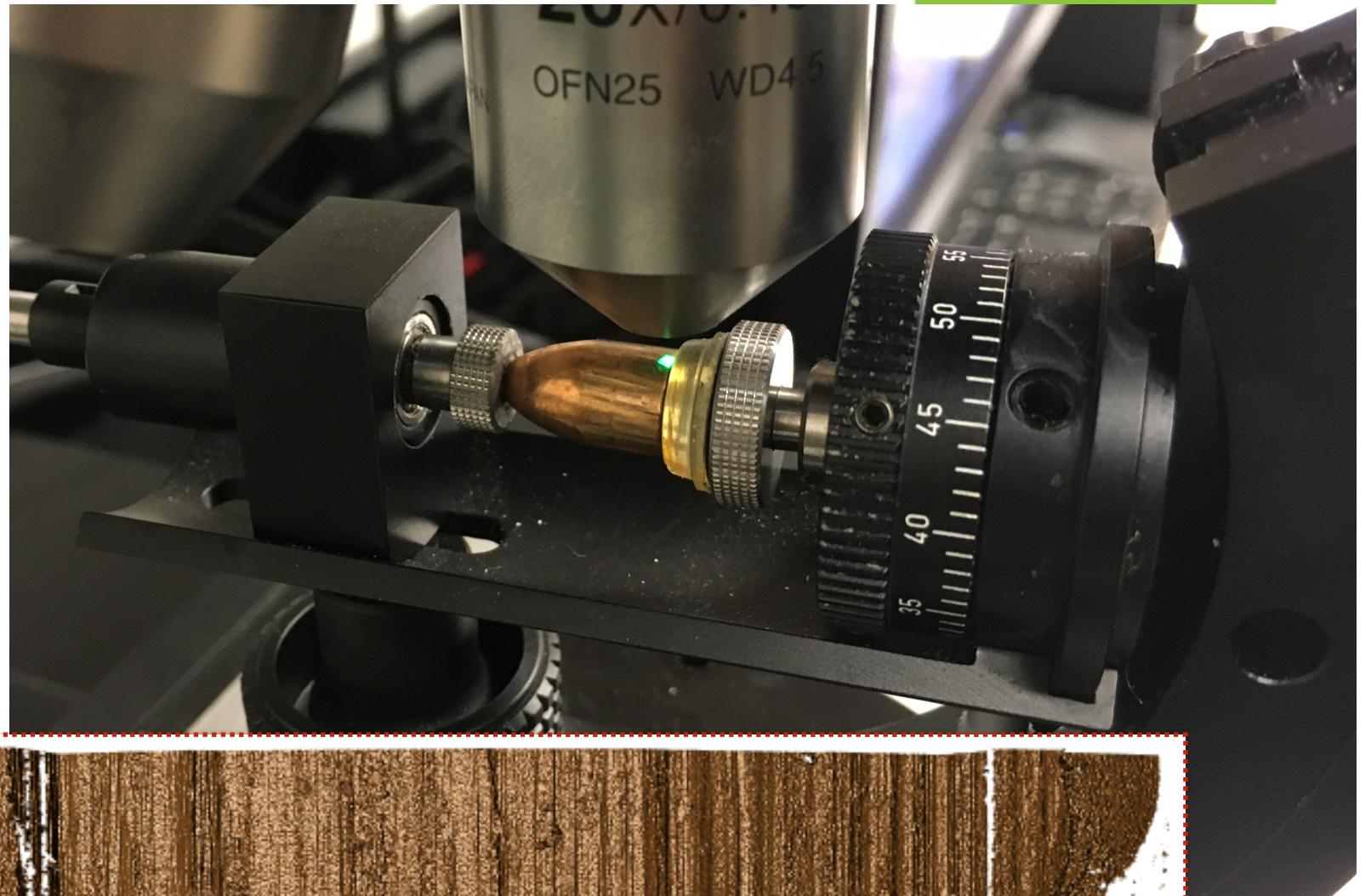
# Data: Microscope Facility

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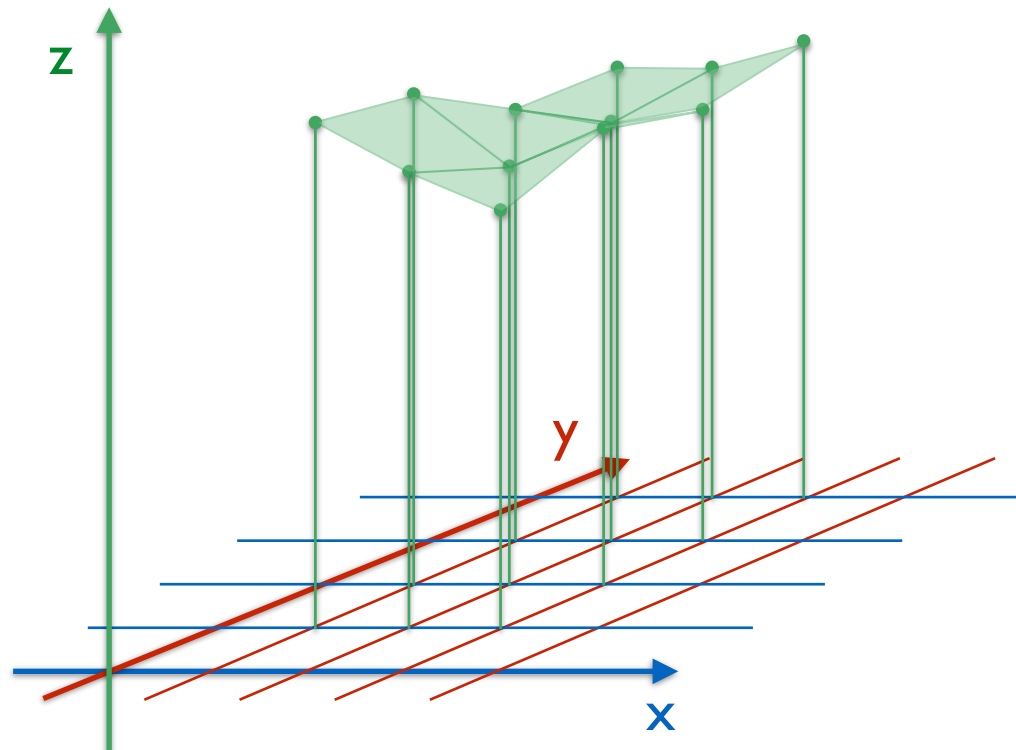
3d topographic images:  
height measurements on  
x-y grid



# Data from CL Microscope

x-y-z files

Data captured on a regular  
grid of  $0.645 \mu m \times 0.645 \mu m$   
Total captured area for each  
land  $\sim 2.2 mm \times 0.6 mm$



x - y - z file

x	y	z
18.705	0.000	-25.221138
19.350	0.000	-25.253155
19.995	0.000	-25.335022
20.640	0.000	-25.418171
21.285	0.000	-25.477917
21.930	0.000	-25.541687
22.575	0.000	-25.673903
23.220	0.000	-25.966341
23.865	0.000	-40.070286
24.510	0.000	-40.407612
25.155	0.000	-40.587063
25.800	0.000	-33.437973
26.445	0.000	-33.691895
27.090	0.000	-39.690674
27.735	0.000	-40.317741

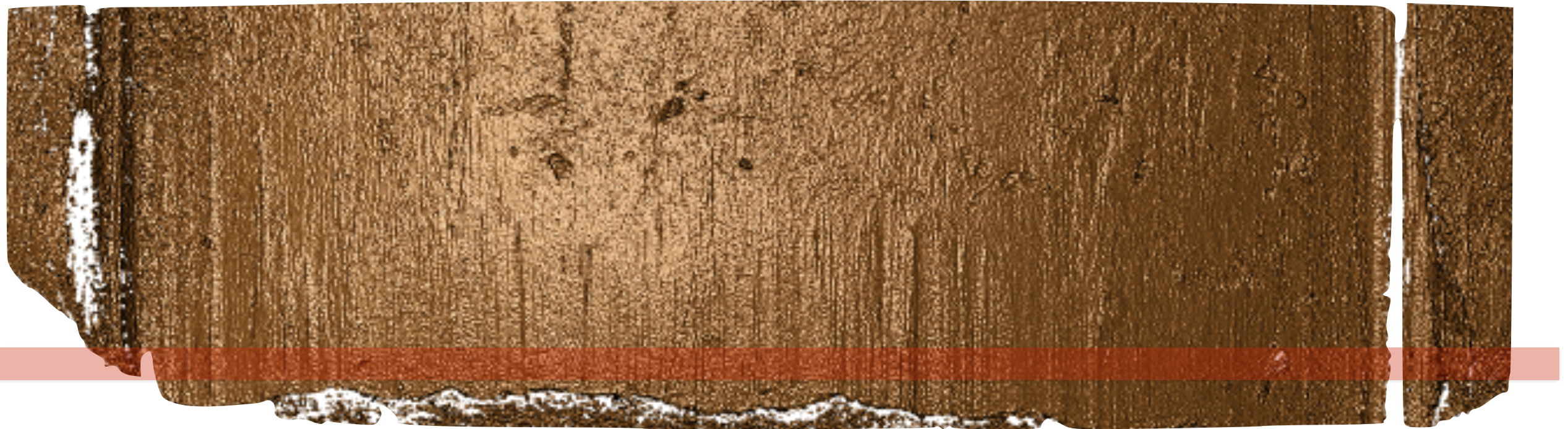
.  
. .  
. .



# Automatic matching score

## Step 1: identify region suitable for matching

land from bullet fired from Smith & Wesson



Region close to heel of bullet

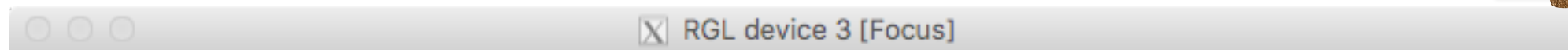
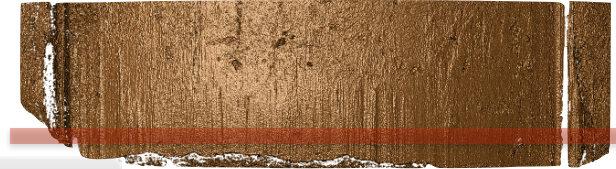
Avoid break-off



# Automatic matching score

Step 1b: from scan to crosscut

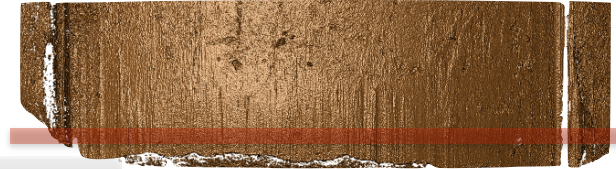
Identify matching region



# Automatic matching score

Step 1b: from scan to crosscut

Identify matching region

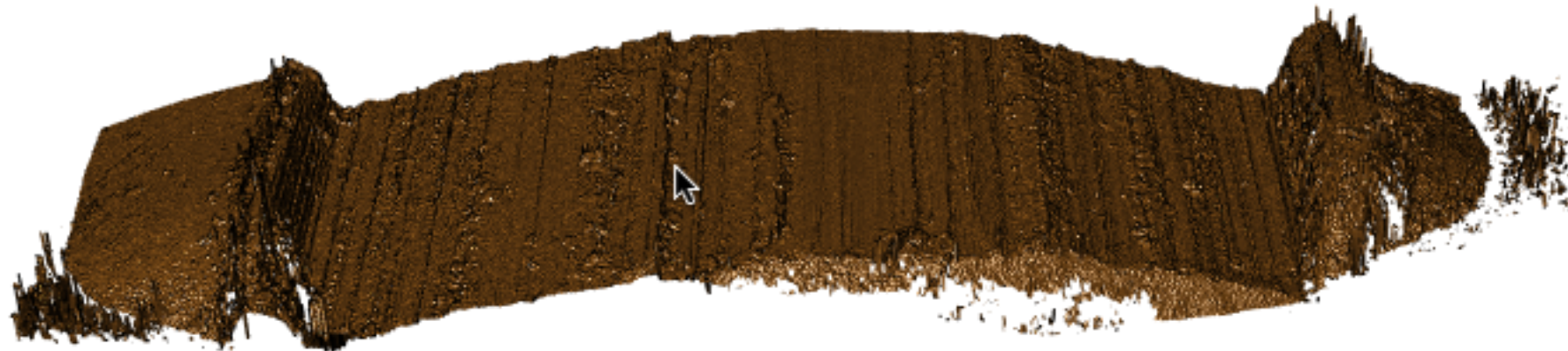
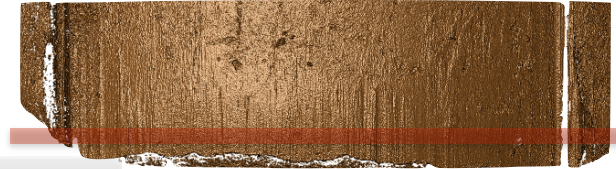




# Automatic matching score

Step 1b: from scan to crosscut

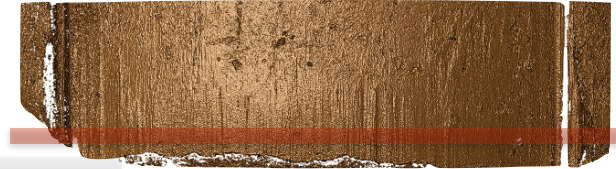
Identify matching region



# Automatic matching score

Step 1b: from scan to crosscut

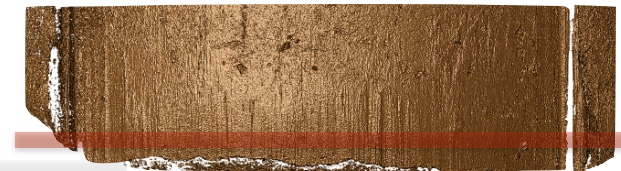
Identify matching region



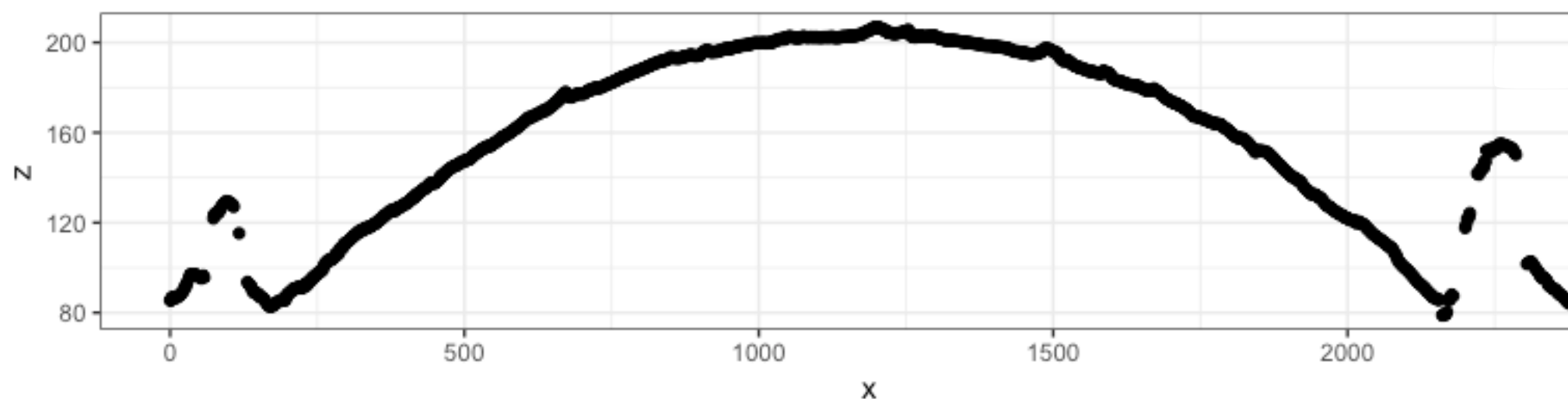
# Automatic matching score

## Step 1b: from scan to crosscut

Identify matching region



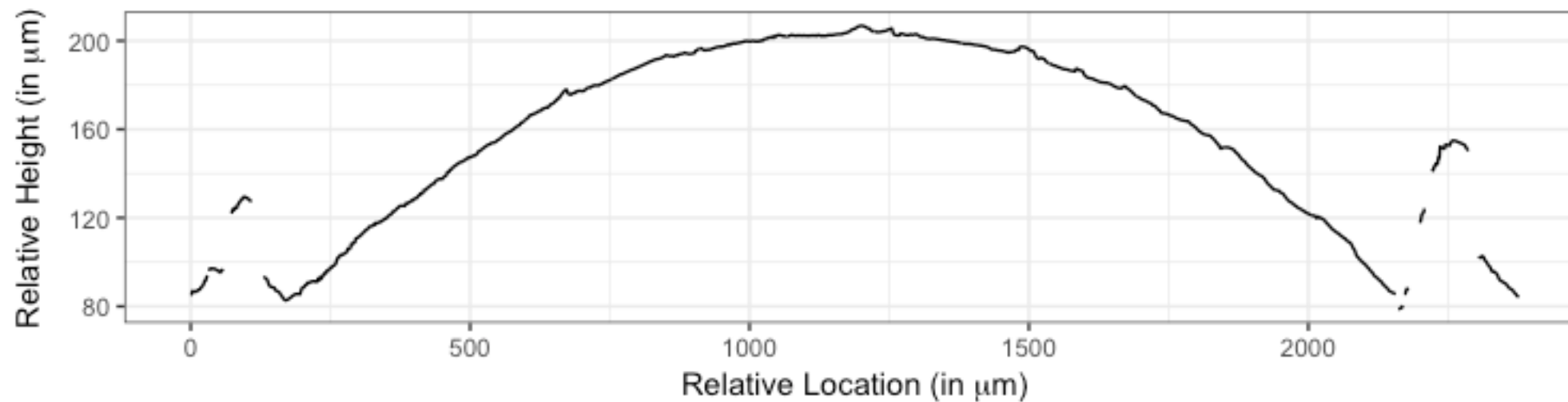
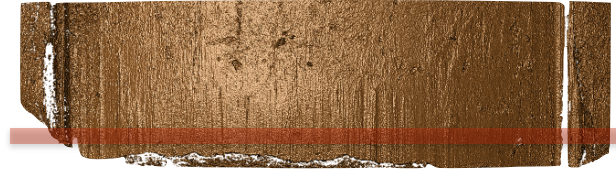
RGL device 3 [Focus]



# Automatic matching score

## Step 2: Identify groove locations

Identify matching region

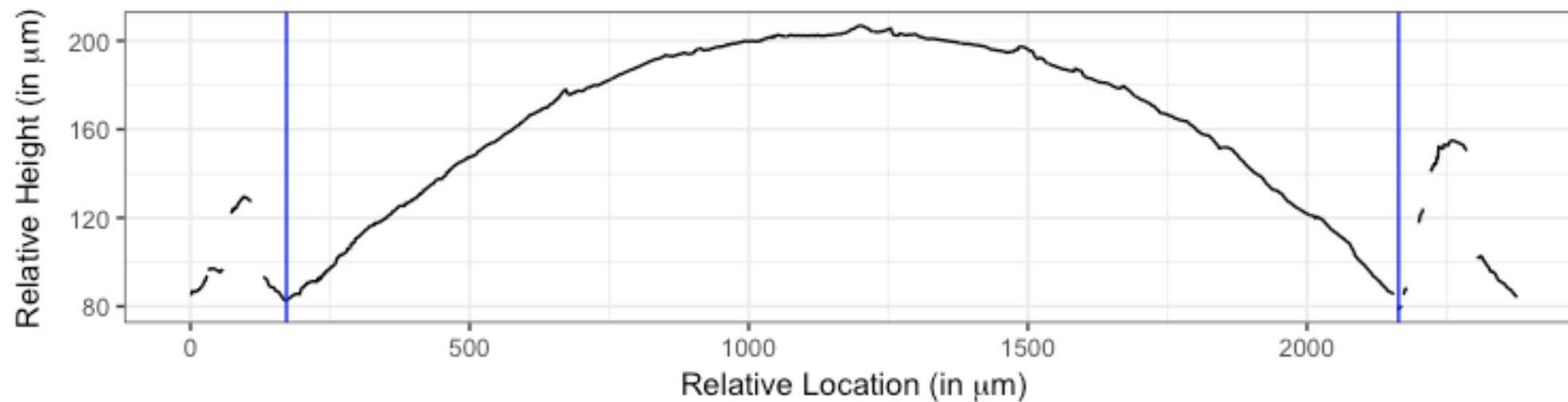
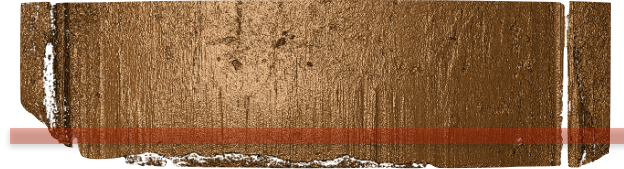


Shoulders (locations outside the grooves) are removed

# Automatic matching score

## Step 2: Identify groove locations

Identify matching region

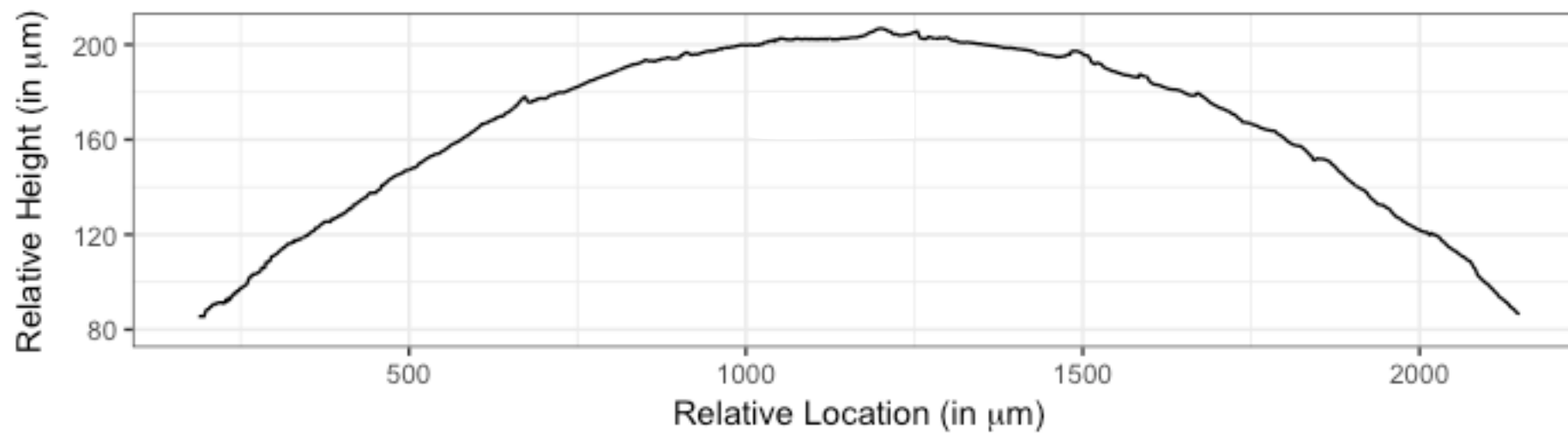


Shoulders (locations outside the grooves) are removed

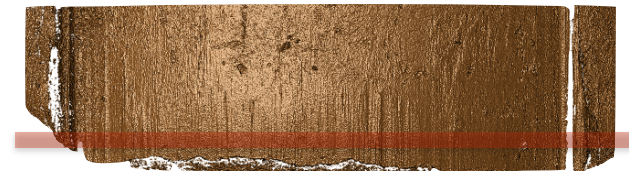


# Automatic matching score

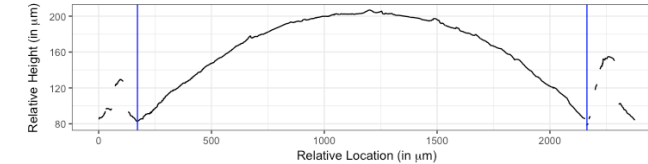
## Step 3: Fit curvature



Identify matching region

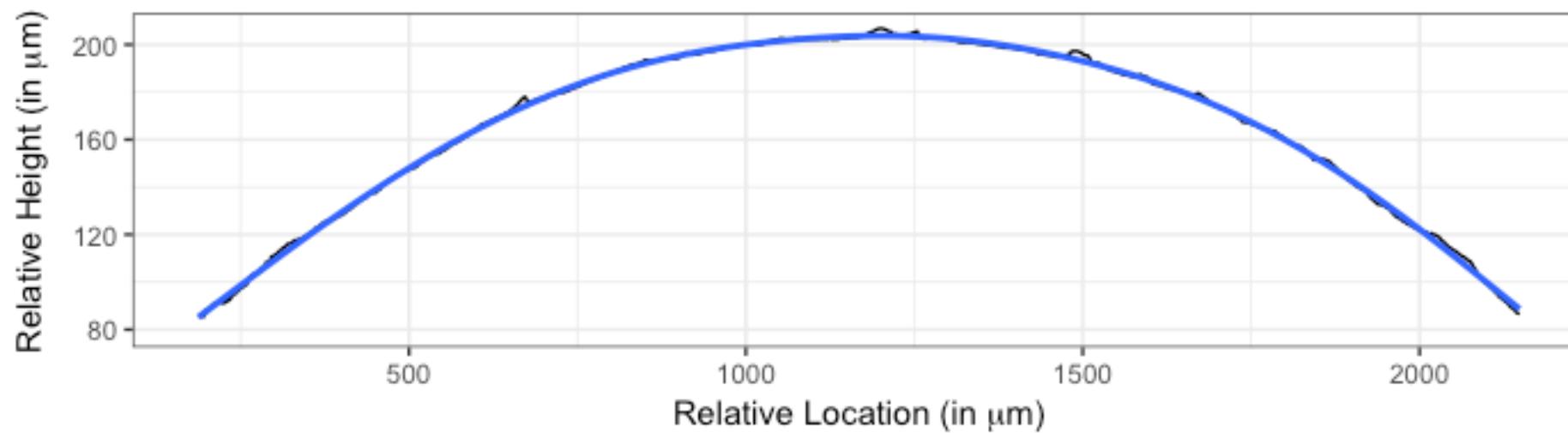


Identify groove locations

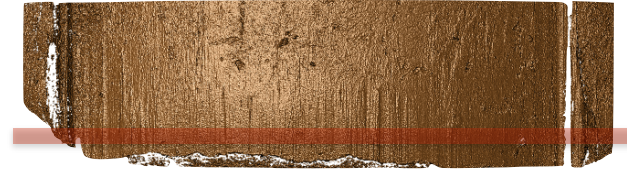


# Automatic matching score

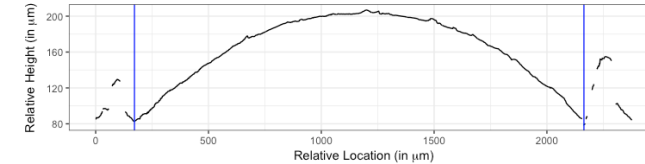
## Step 3: Fit curvature



Identify matching region

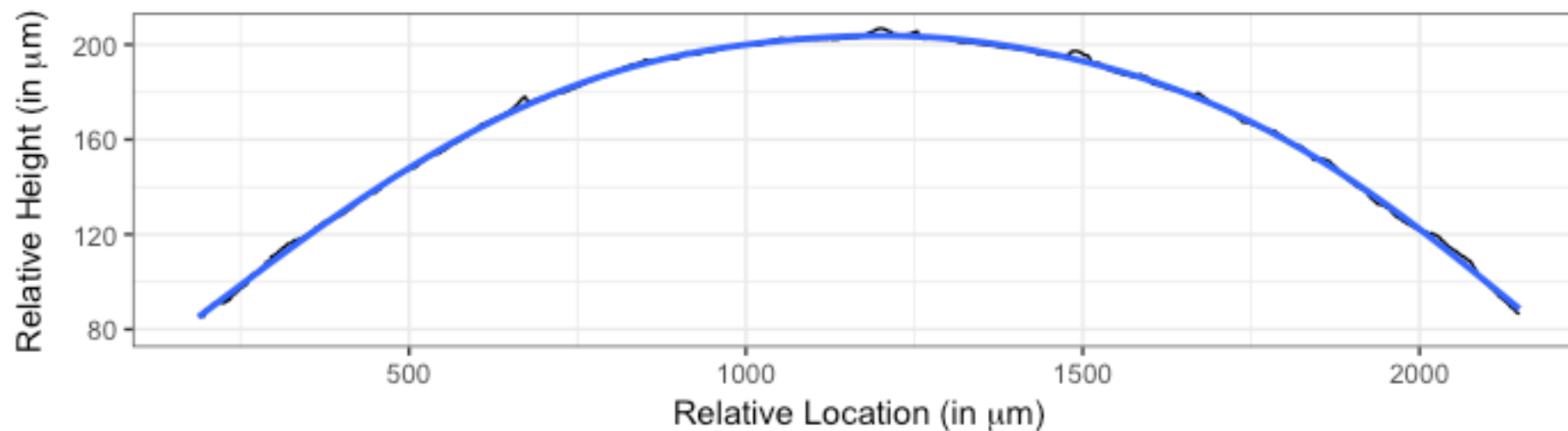


Identify groove locations

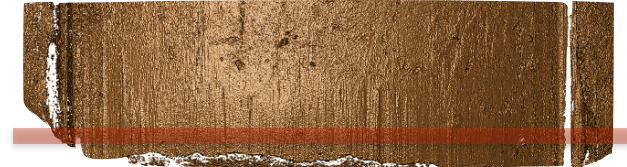


# Automatic matching score

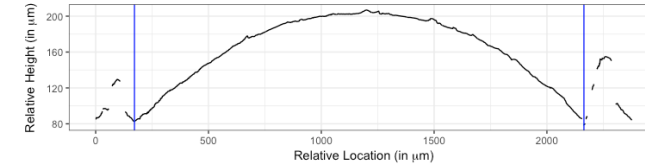
## Step 3: Fit curvature & get signature



Identify matching region



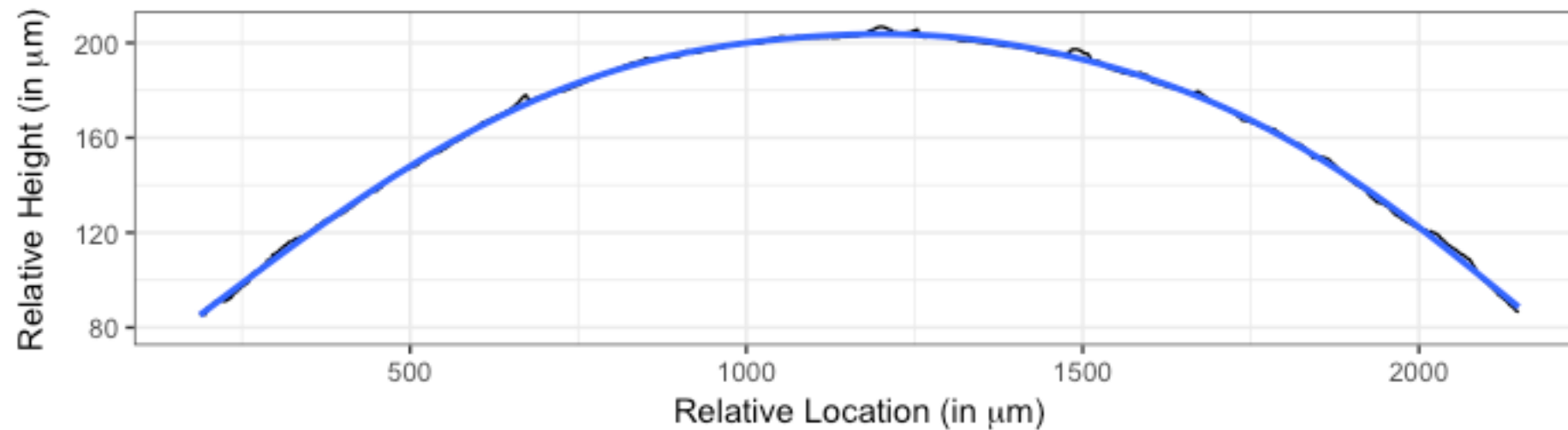
Identify groove locations



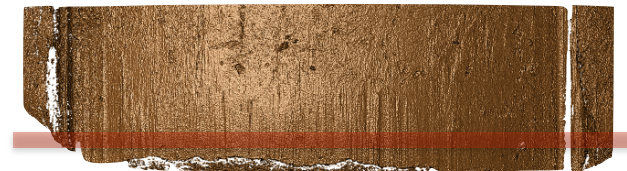


# Automatic matching score

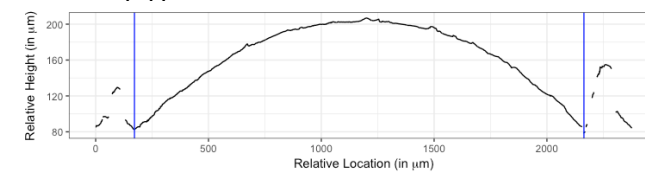
## Step 3: Fit curvature & get signature



Identify matching region

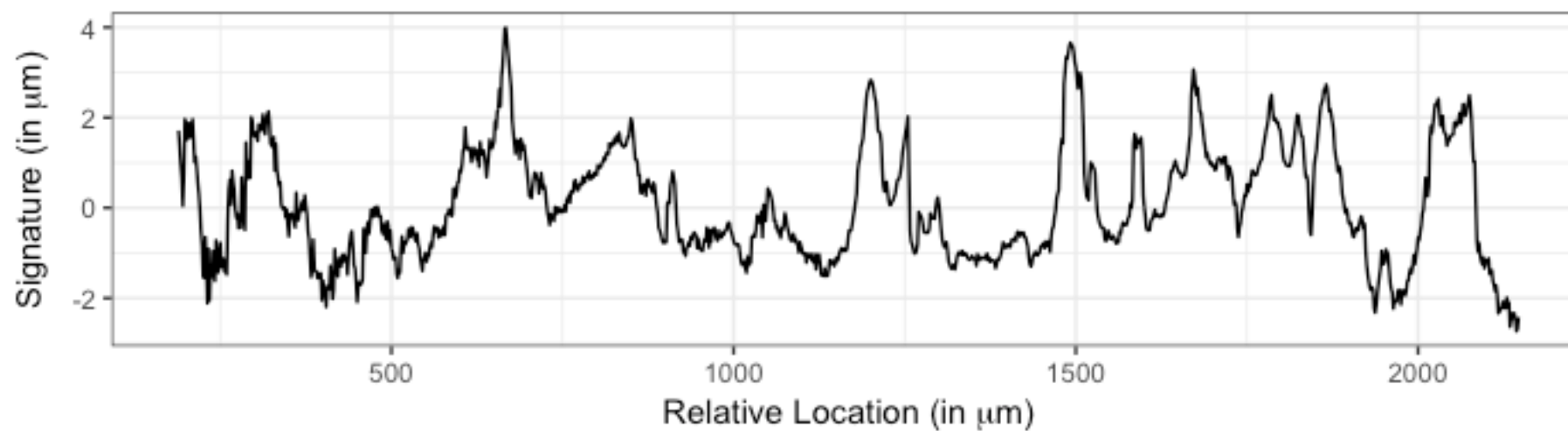
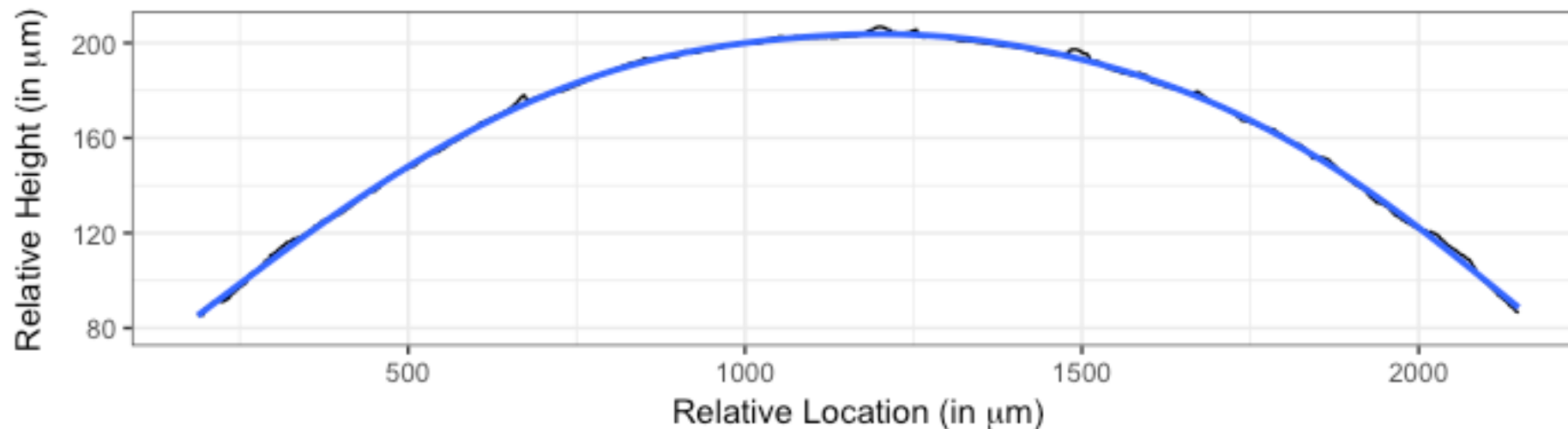


Identify groove locations

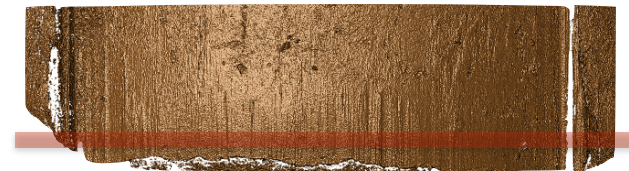


# Automatic matching score

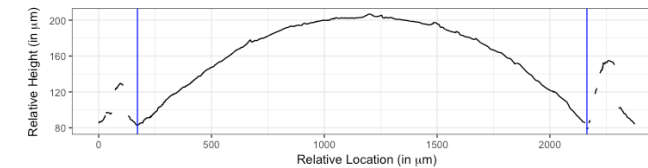
## Step 3: Fit curvature & get signature



Identify matching region

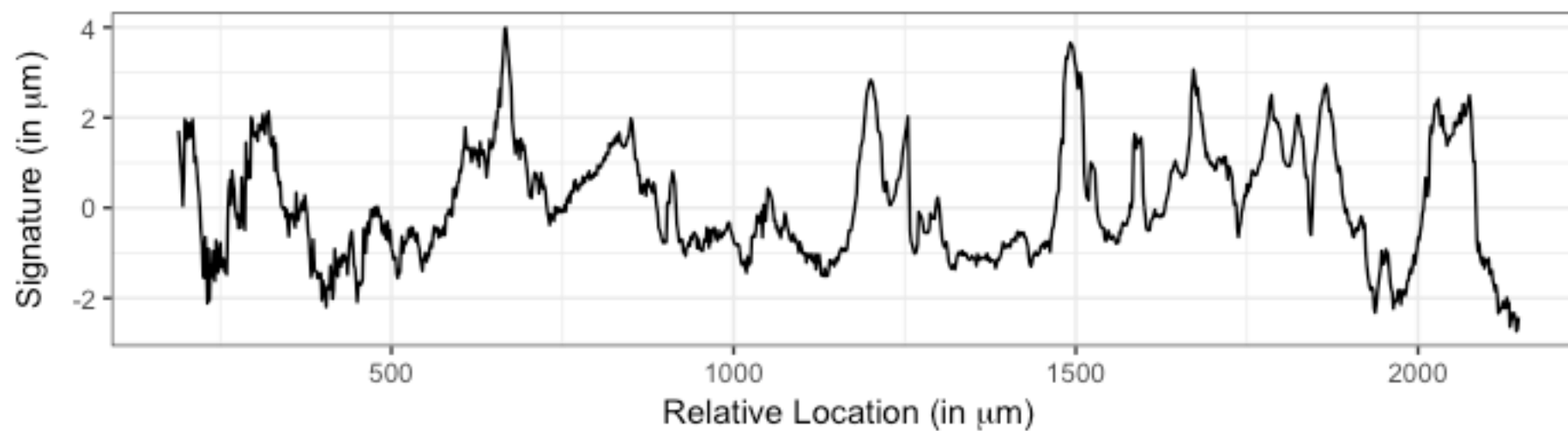


Identify groove locations

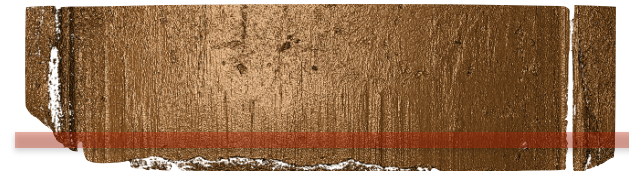


# Automatic matching score

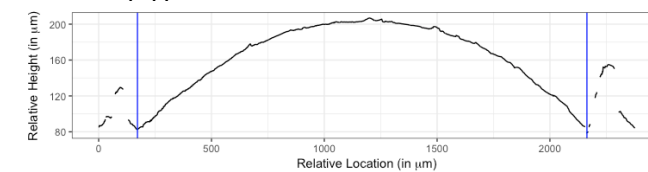
## Step 4: Align signatures



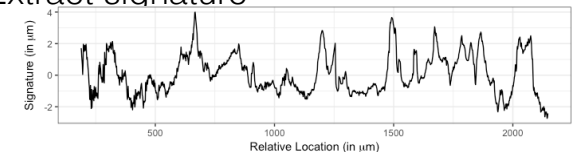
Identify matching region



Identify groove locations

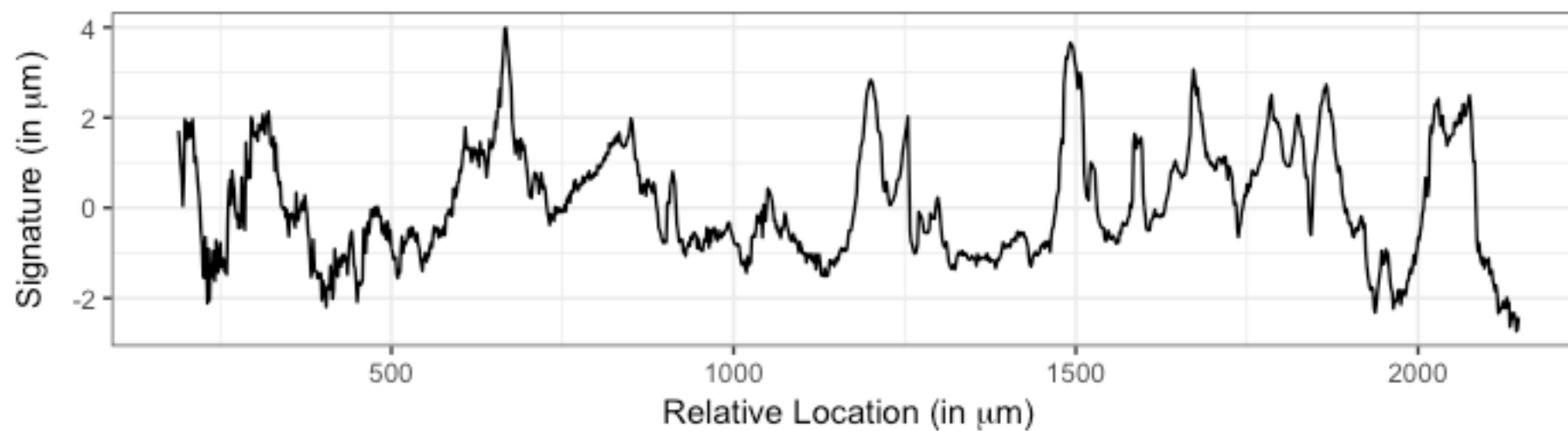


Extract signature

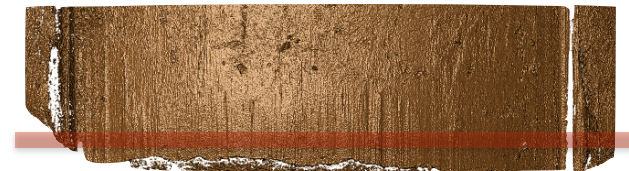


# Automatic matching score

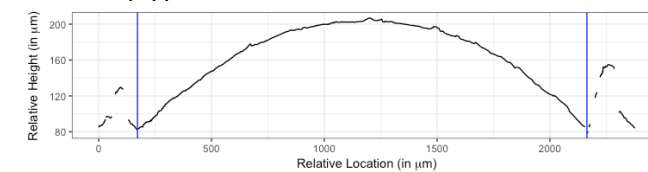
## Step 4: Align signatures



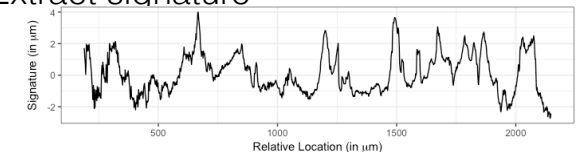
Identify matching region



Identify groove locations



Extract signature

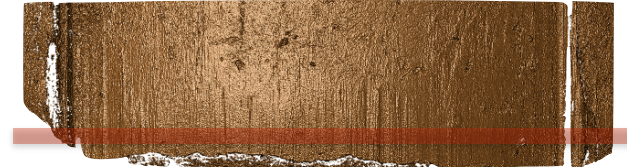


# Automatic matching score

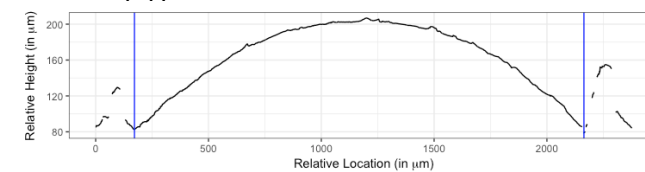
## Step 4: Align signatures



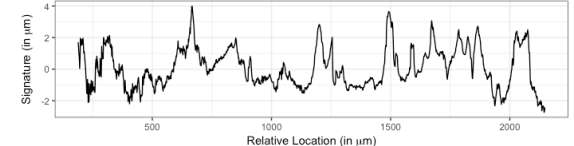
Identify matching region



Identify groove locations



Extract signature

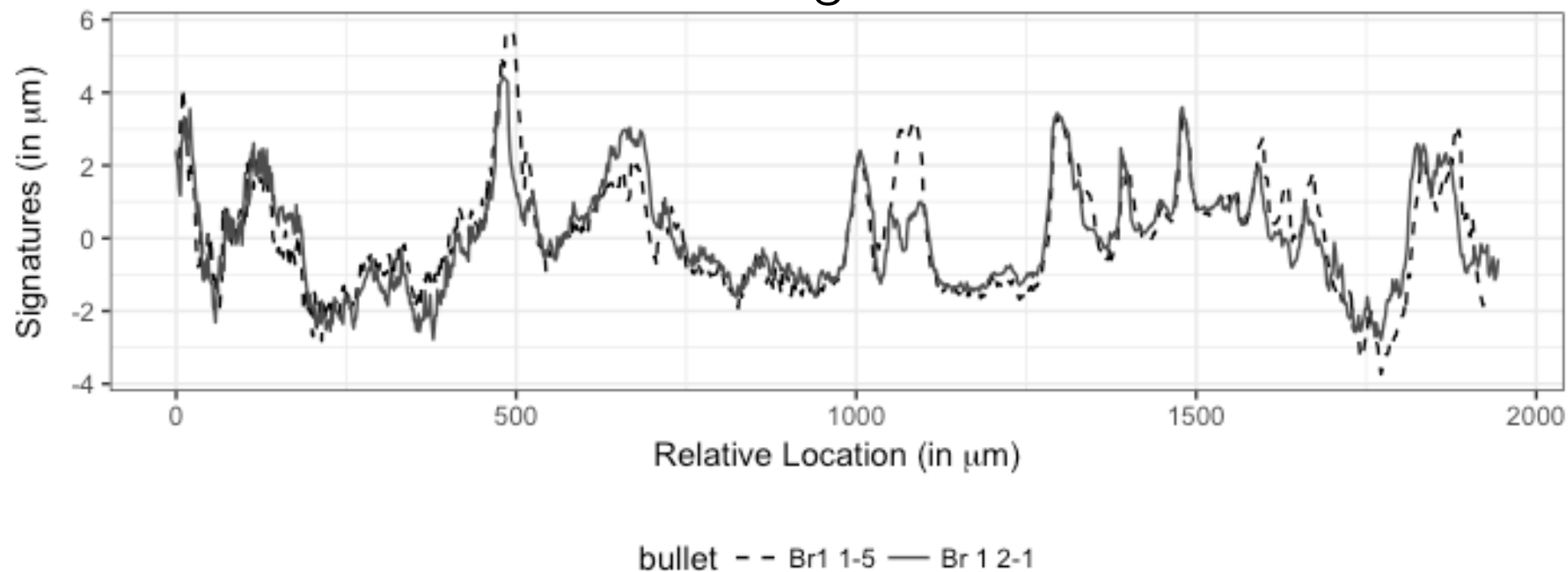




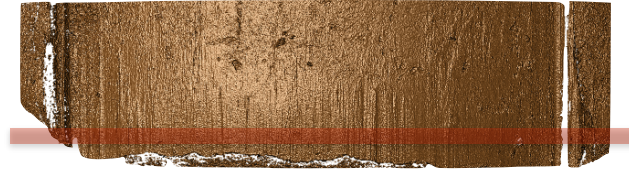
# Automatic matching score

## Step 4: Align signatures

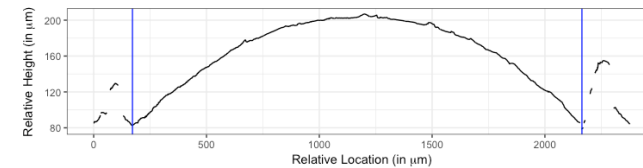
Horizontal shifts to find best alignment



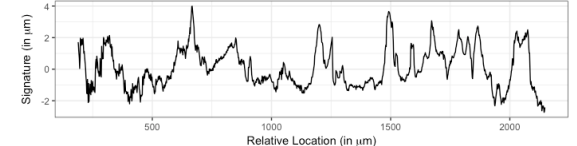
Identify matching region



Identify groove locations



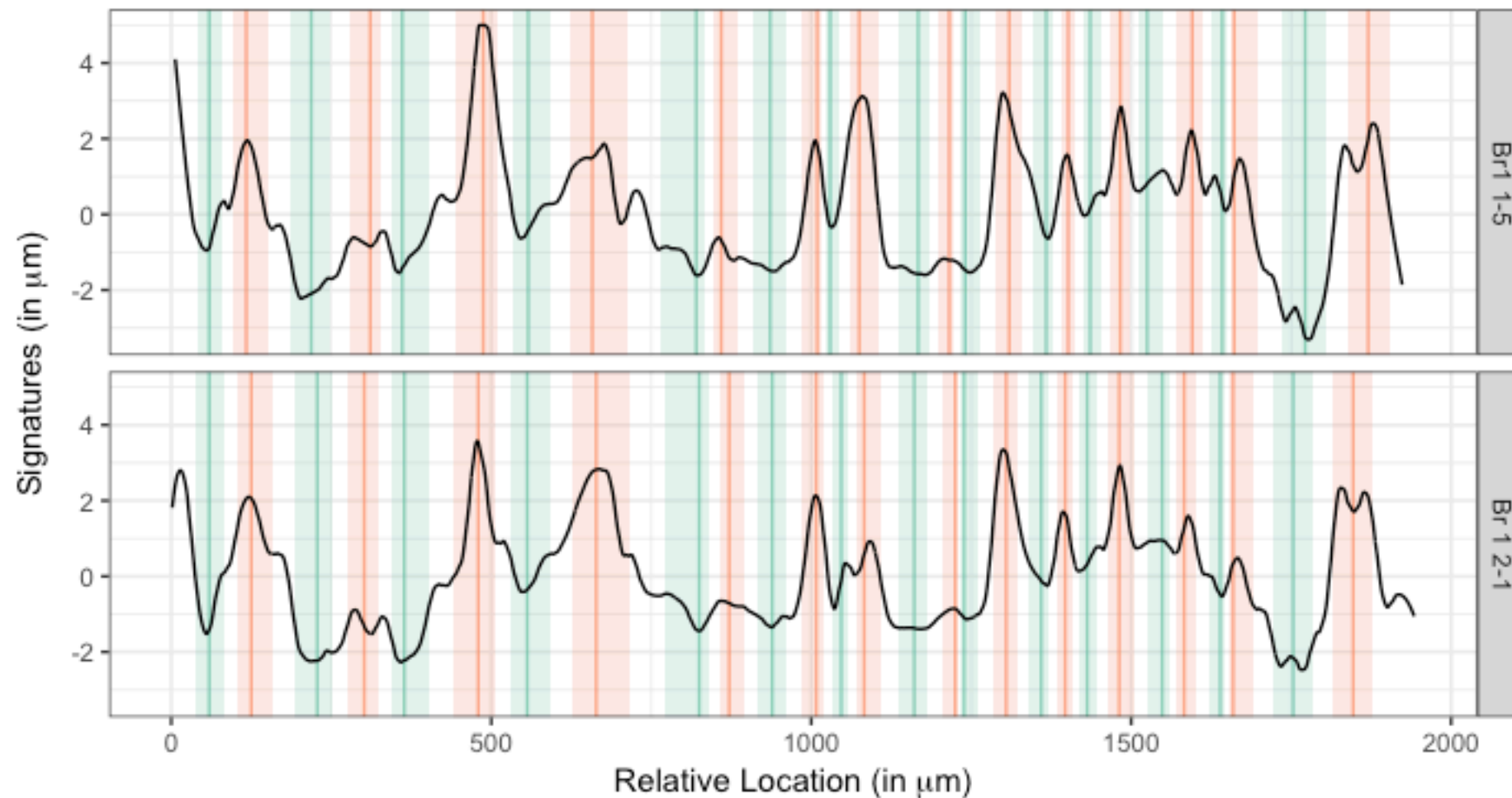
Extract signature



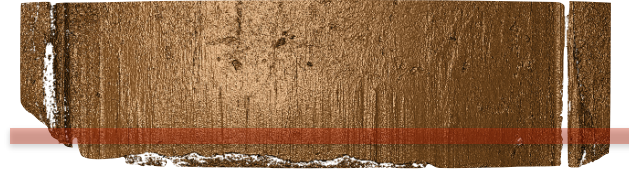
# Automatic matching score

## Step 5: Identify Peaks & Valleys

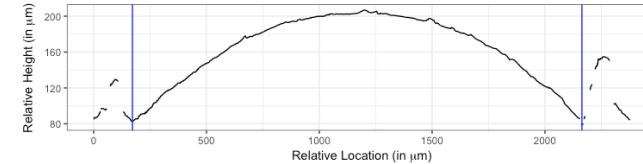
Peaks and valleys in the same locations of two lands are matching striae on the scans



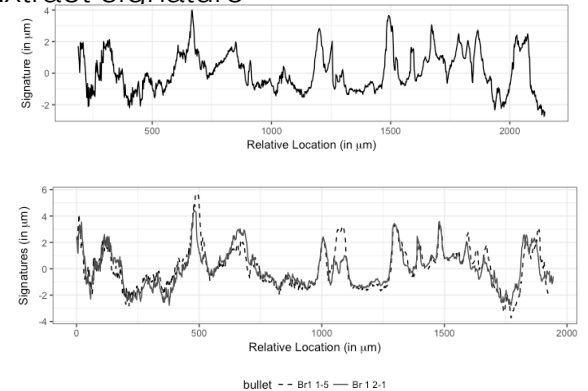
Identify matching region



Identify groove locations



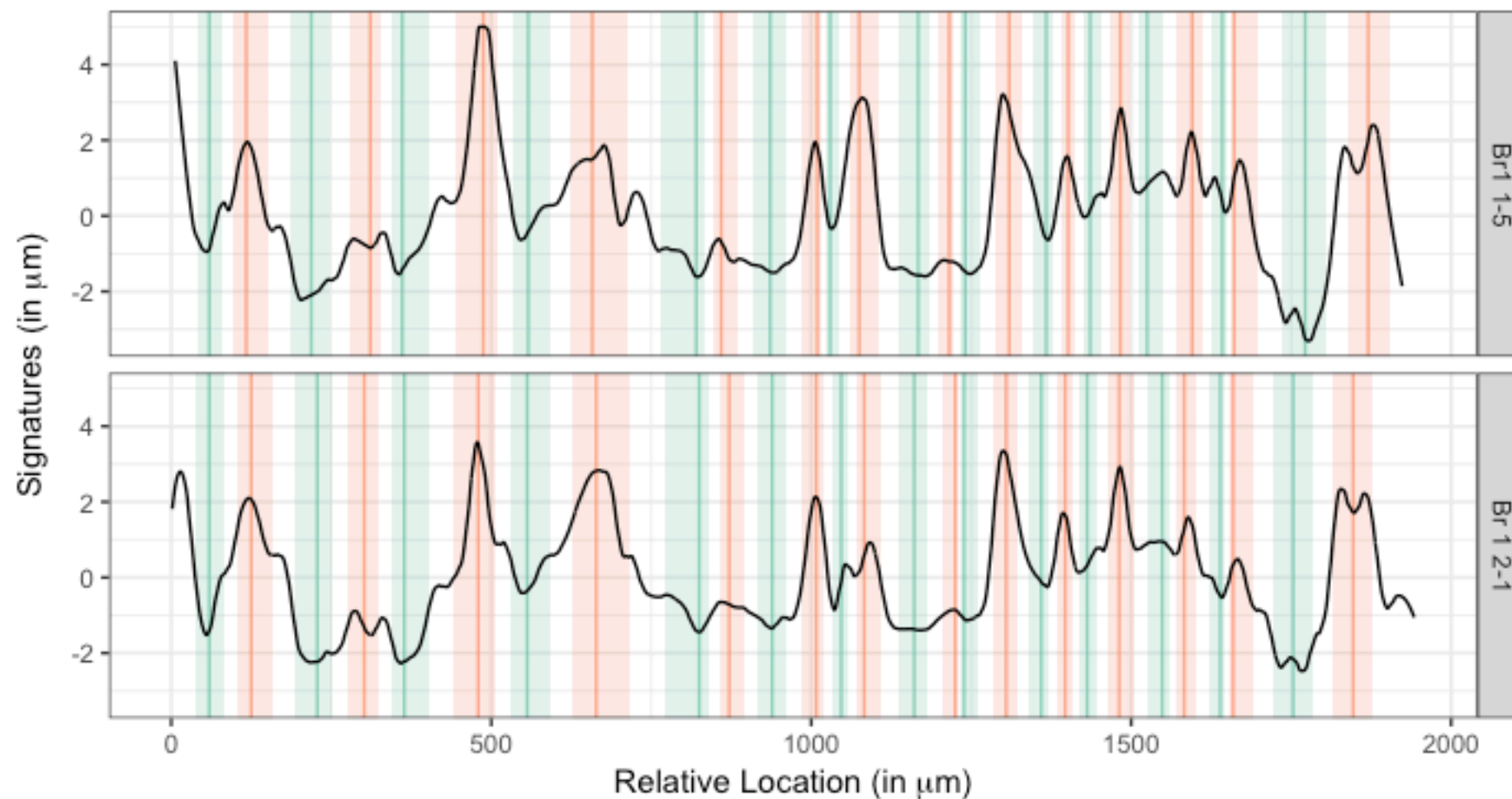
Extract signature



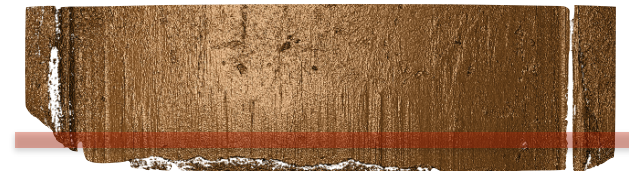
# Automatic matching score

## Step 6: Extract features

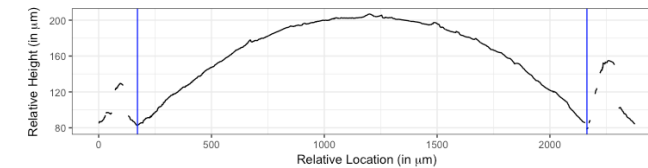
Feature should distinguish between a match and a non-match



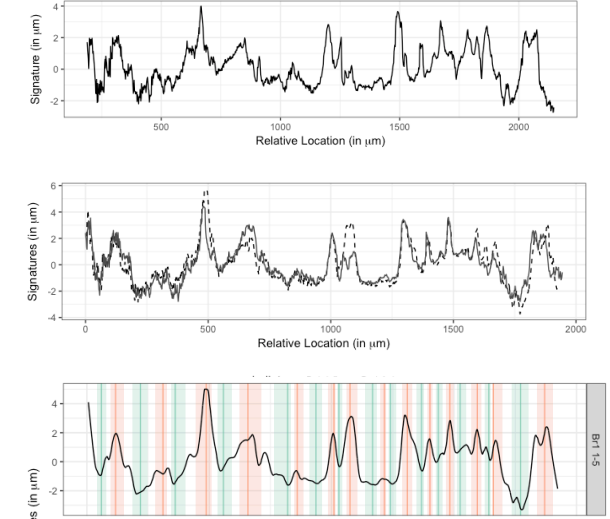
Identify matching region



Identify groove locations



Extract signature

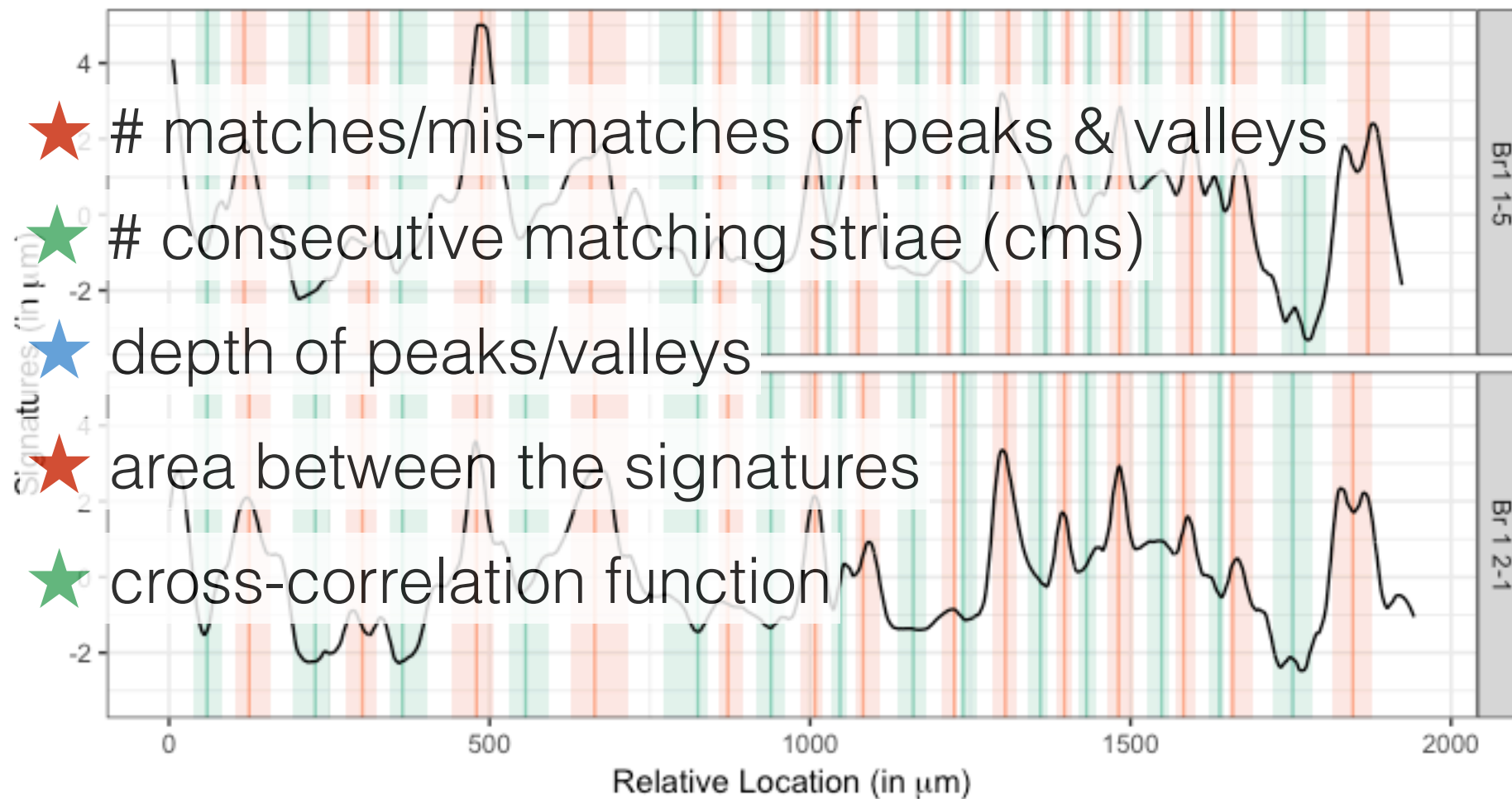




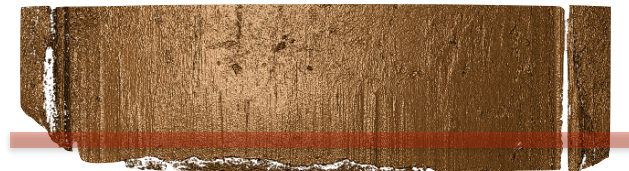
# Automatic matching score

## Step 6: Extract features

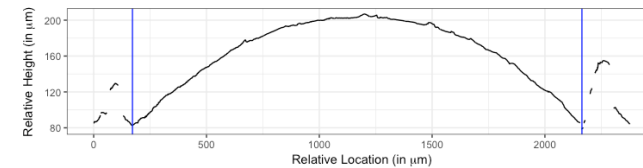
Feature should distinguish between a match and a non-match



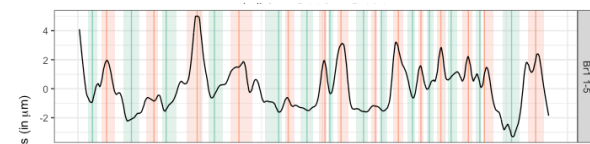
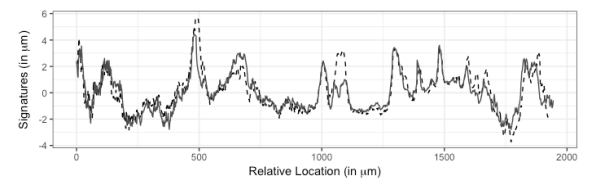
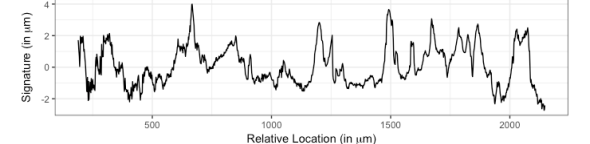
Identify matching region



Identify groove locations



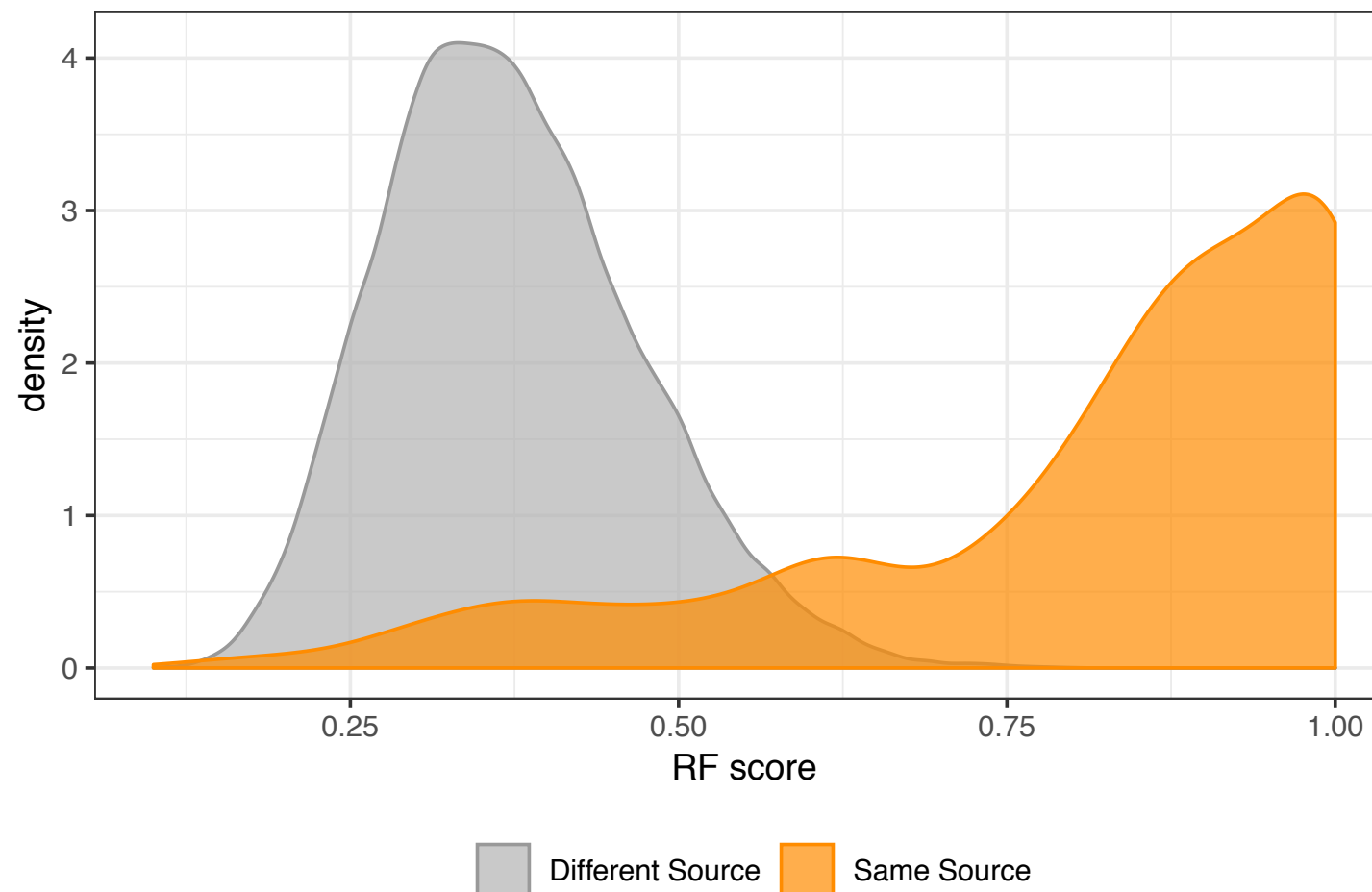
Extract signature



# Automatic matching score

## Step 7: Assign Score (Random Forest Model)

- ★ Matching score between 0 and 1
- ★ Higher score indicates more similarity between two lands

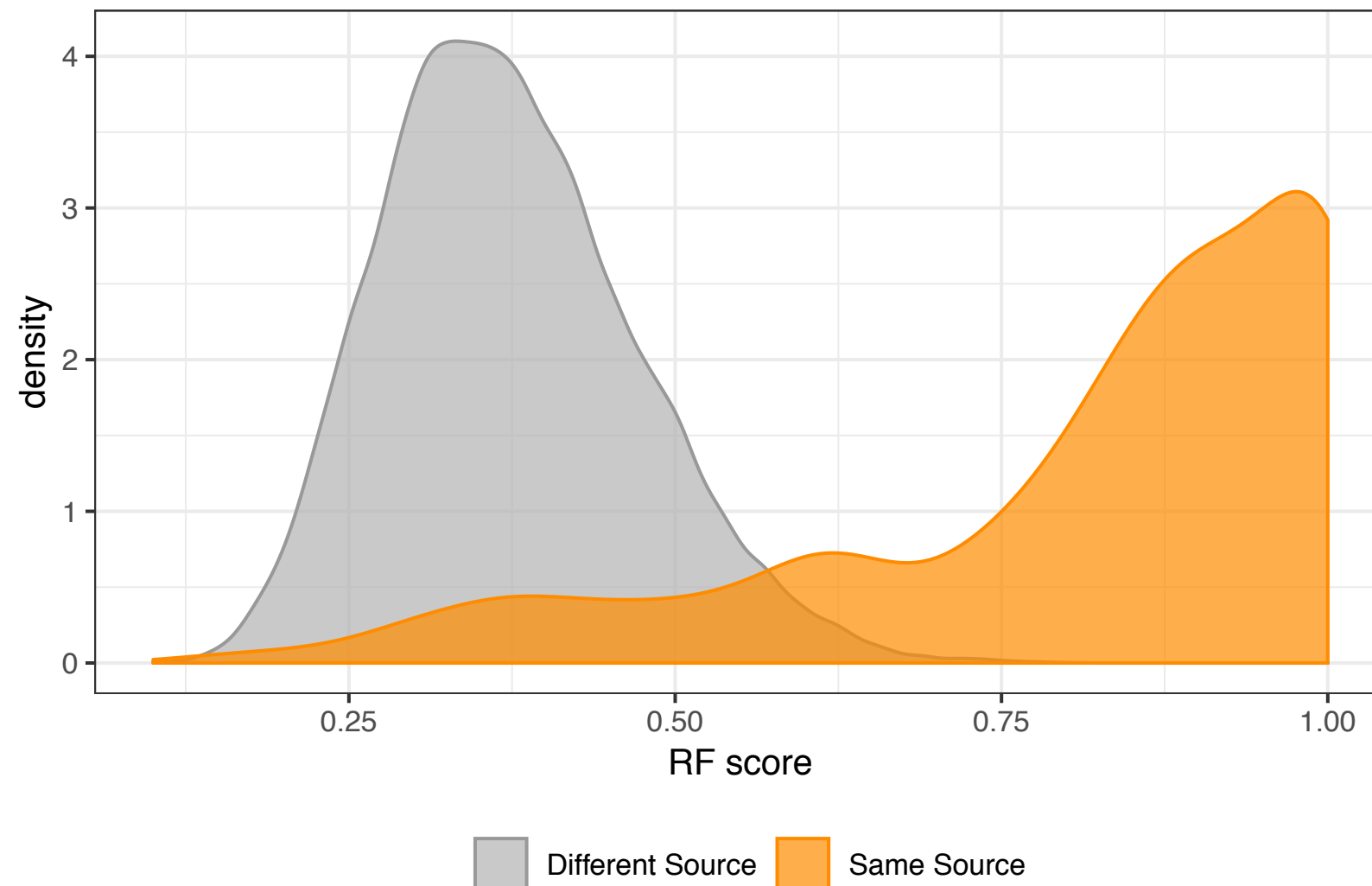


*Automatic matching of  
bullet land impressions,*  
Annals of Applied  
Statistics,  
Eric Riemer Hare, Heike  
Hofmann, and Alicia  
Carriquiry

*Algorithmic approaches to  
match degraded land  
impressions*  
Eric Hare; Heike Hofmann;  
Alicia Carriquiry  
Law, Probability and Risk,  
Volume 16, Issue 4, 1  
December 2017, 203–221,  
<https://doi.org/10.1093/lpr/mgx018>

# What does it mean?

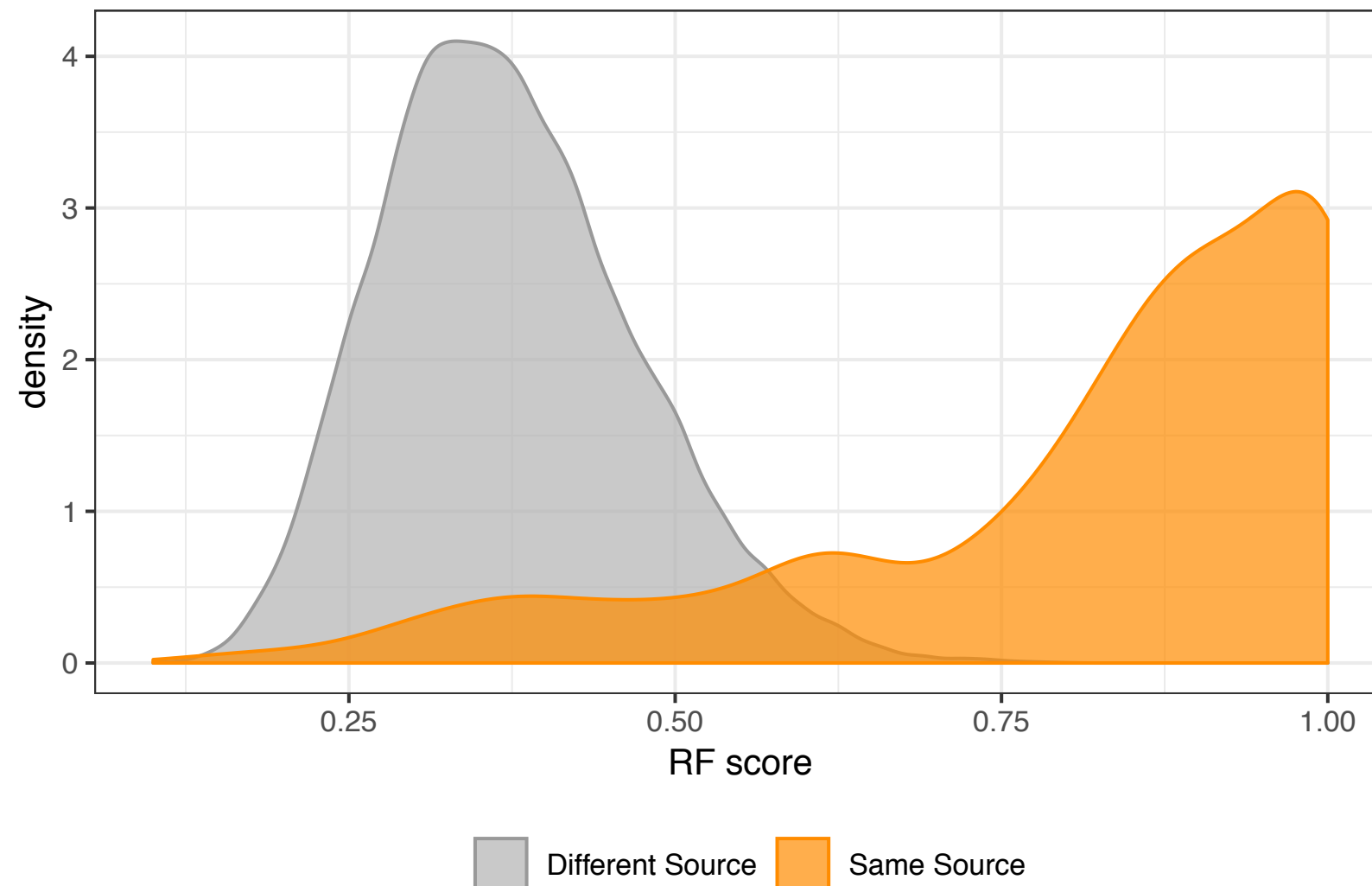
- ★ let's assume that two bullet (land)s are compared, the model spits out a number ...



# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

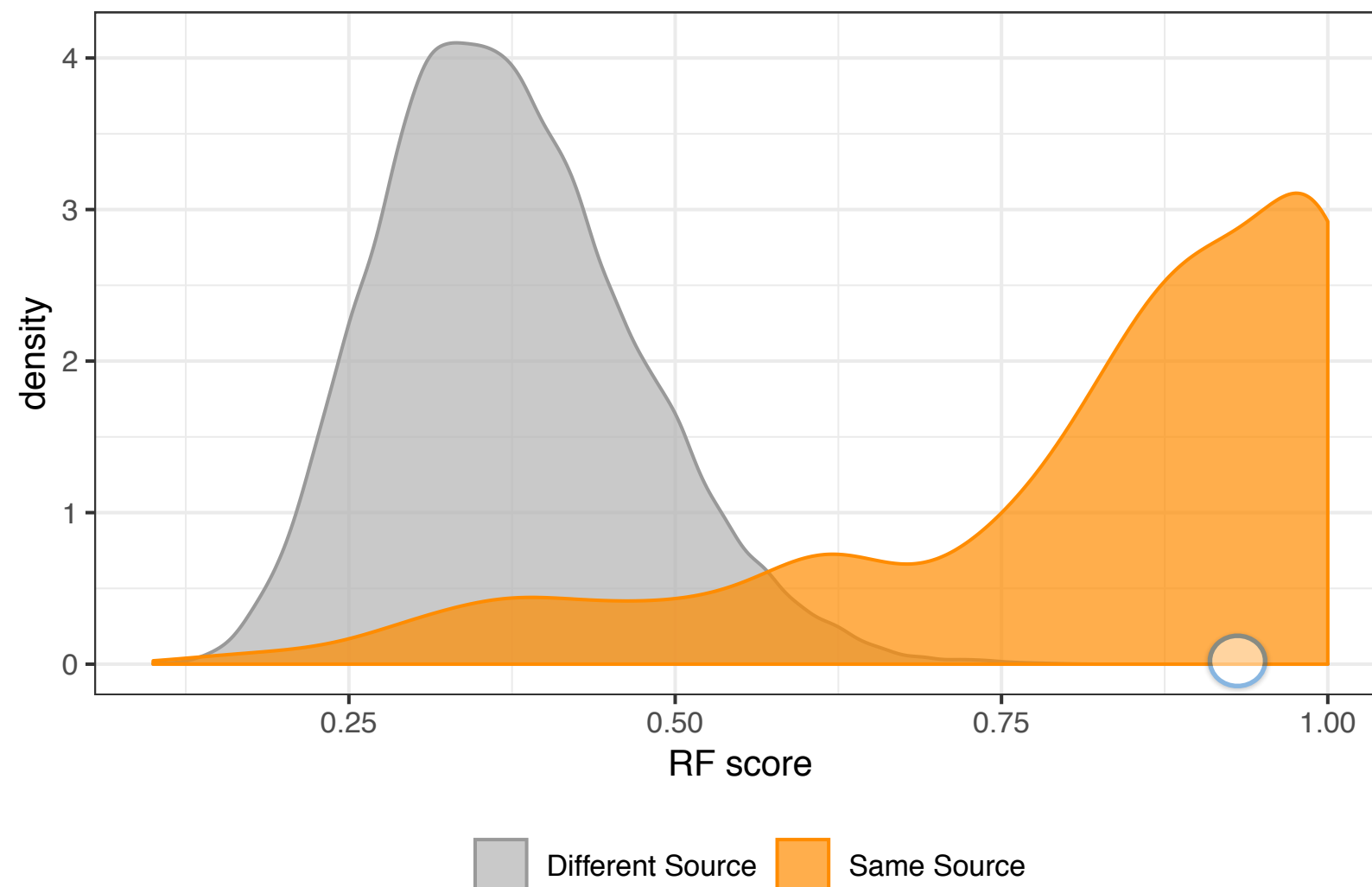
○ 0.91



# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

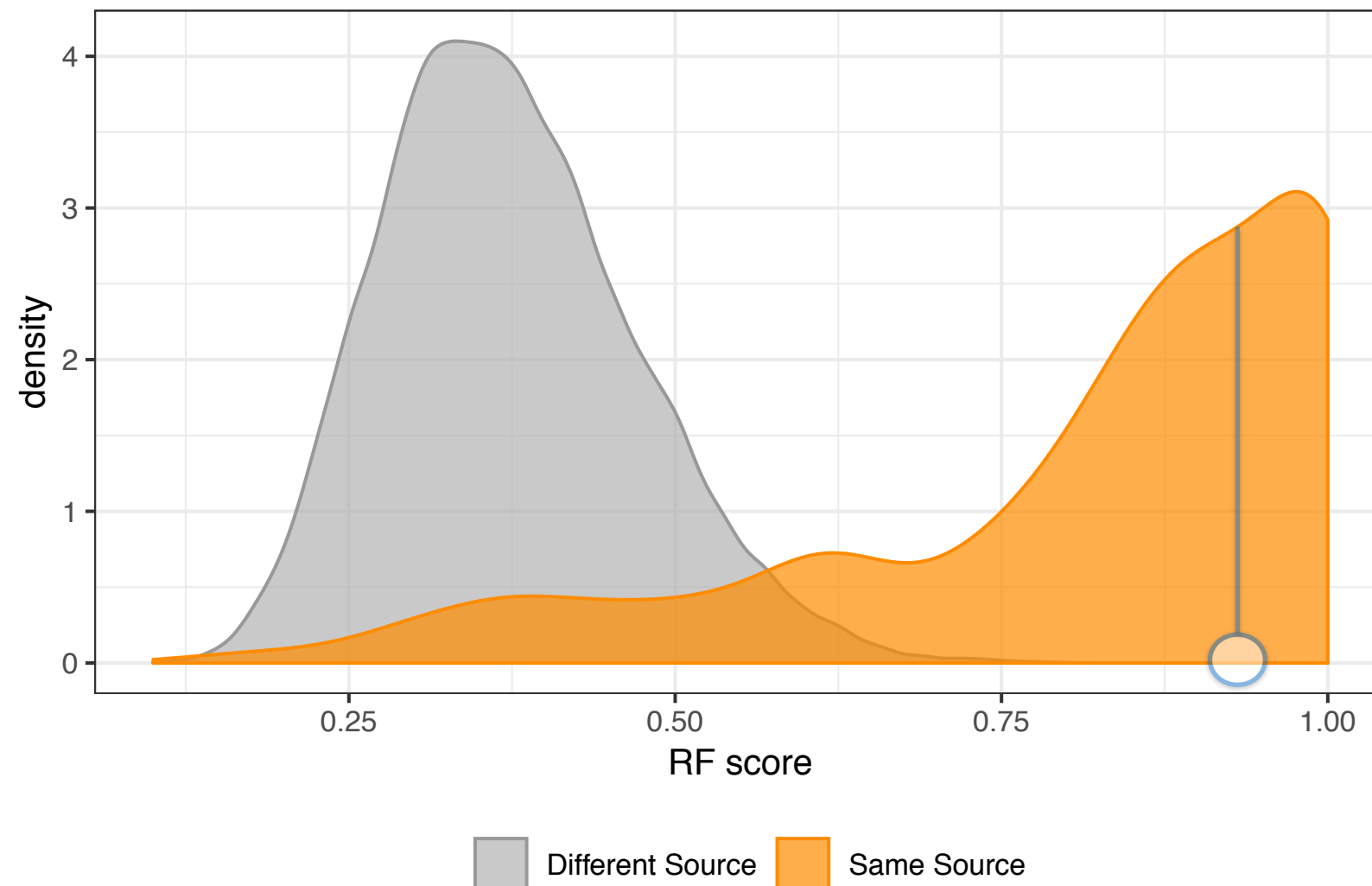
○ 0.91



# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

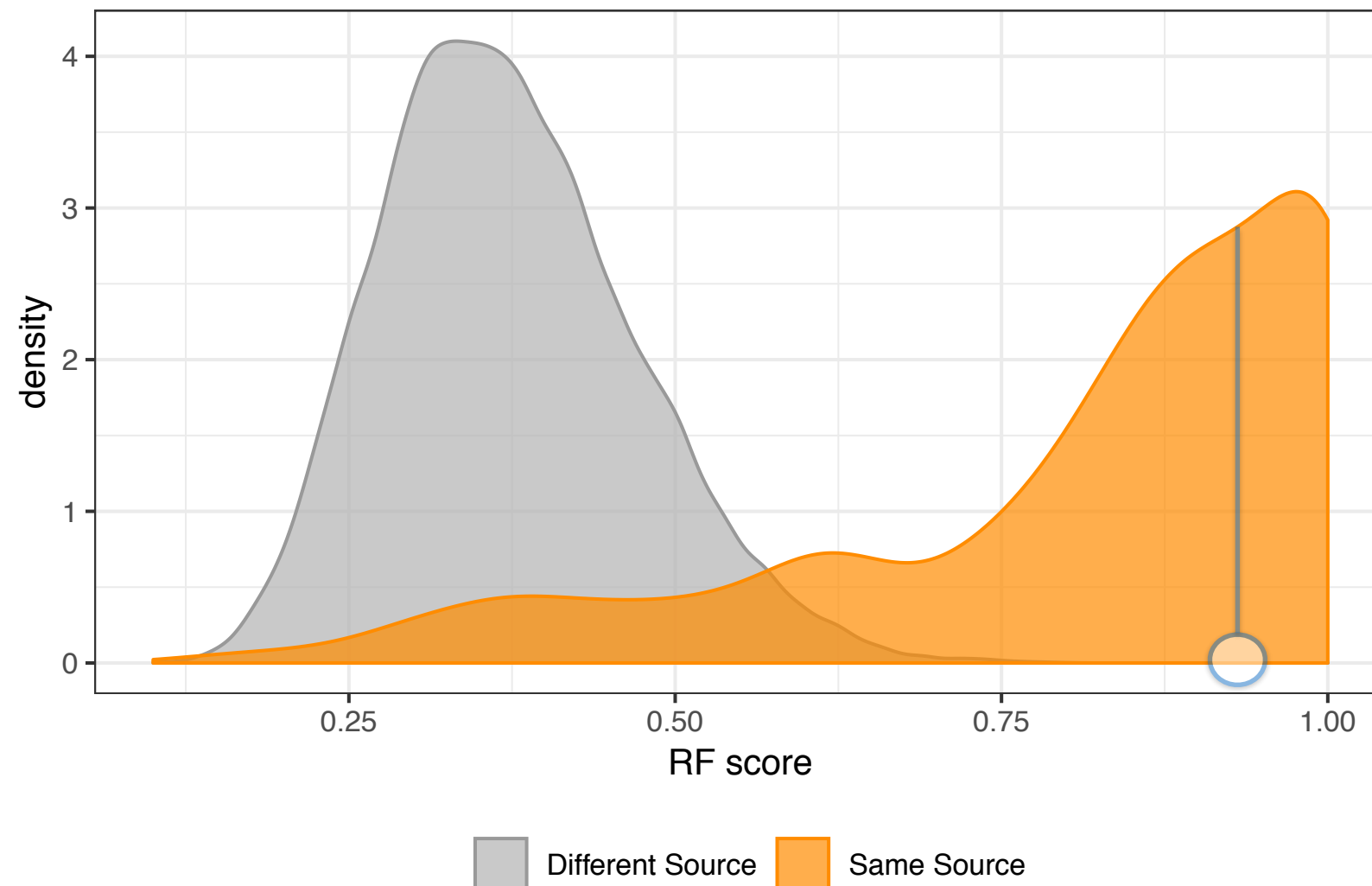


# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

2.77

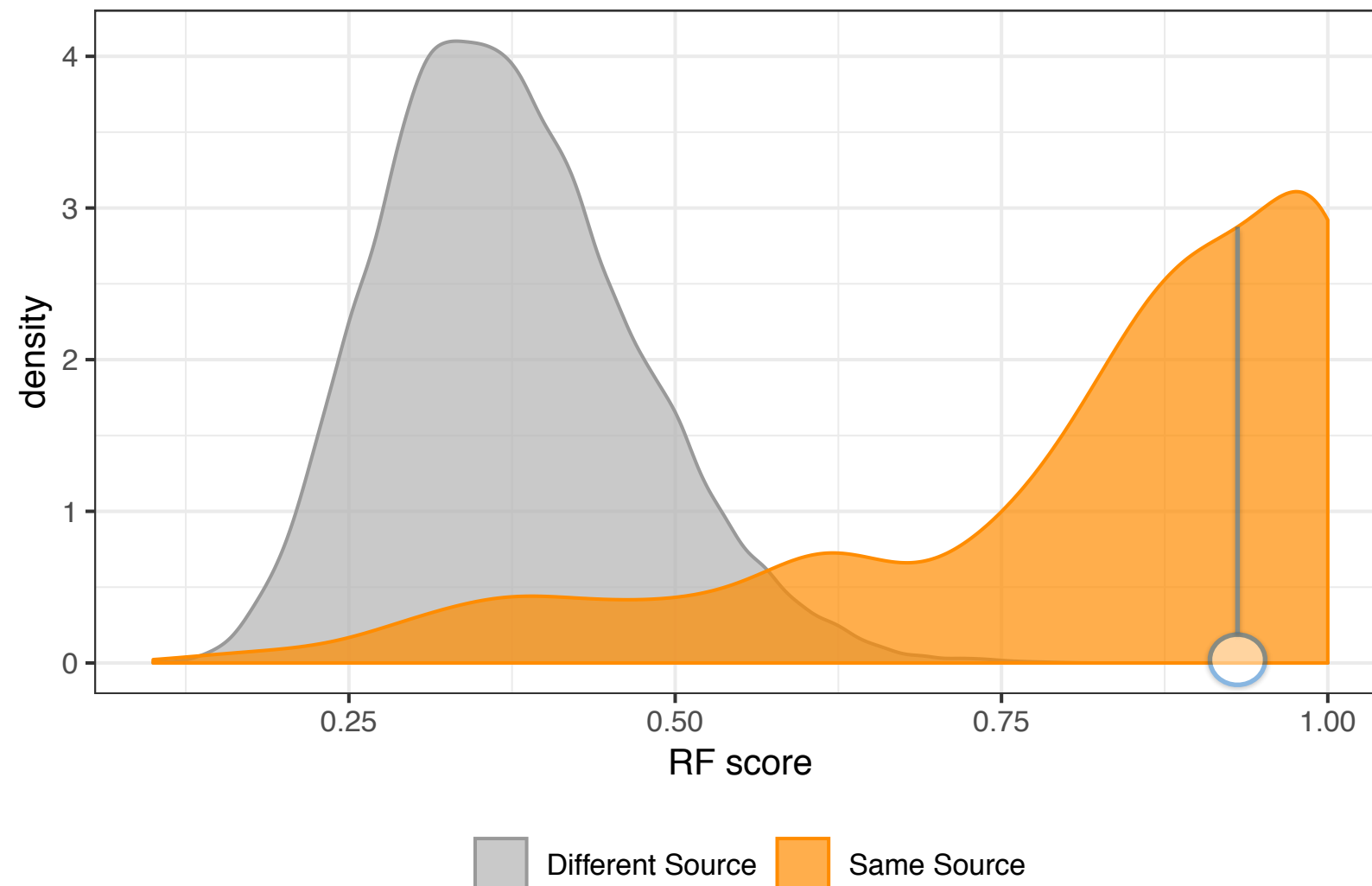


# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

2.77  $1.6 \times 10^{-16}$



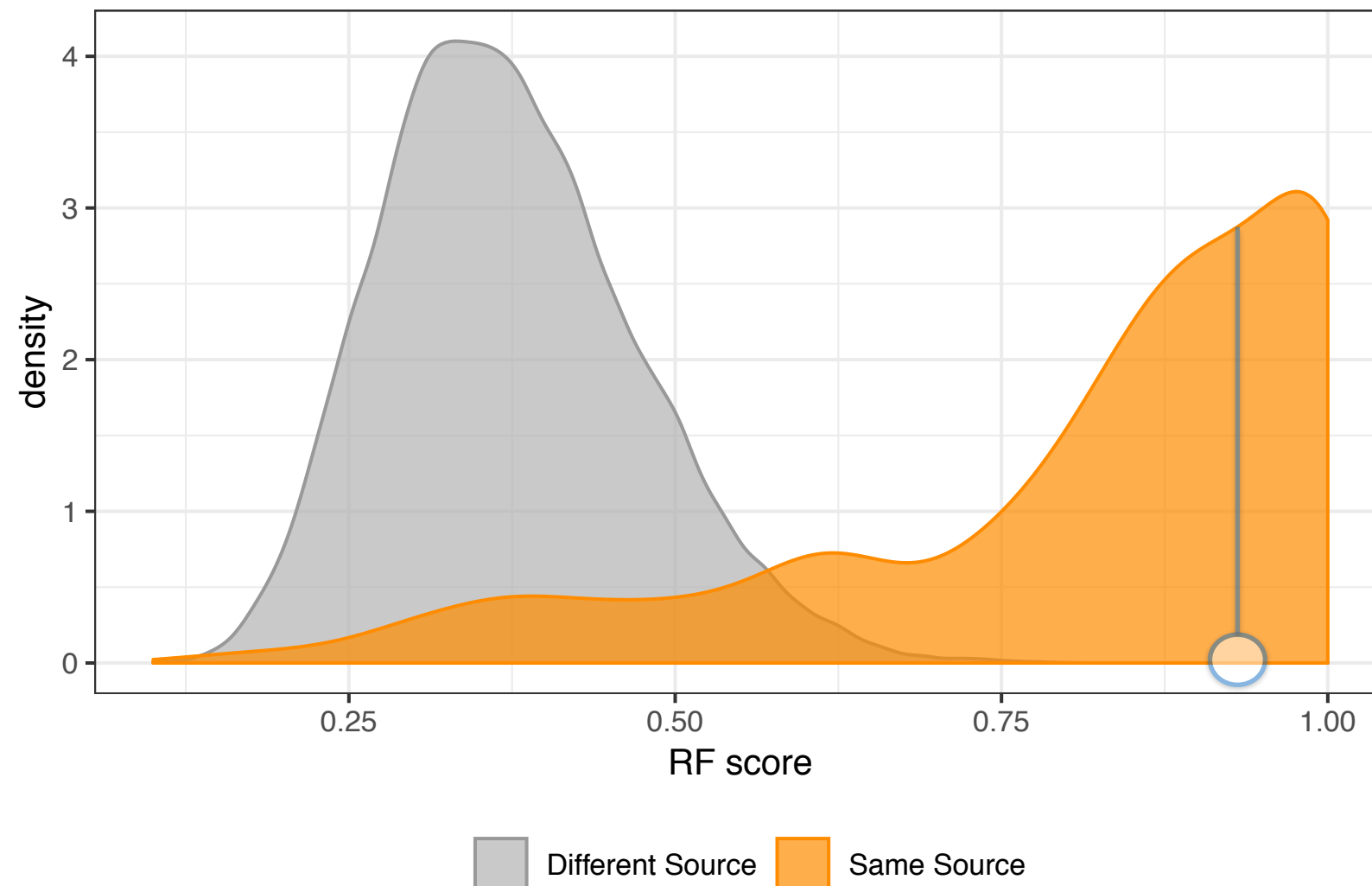


# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

$$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$$

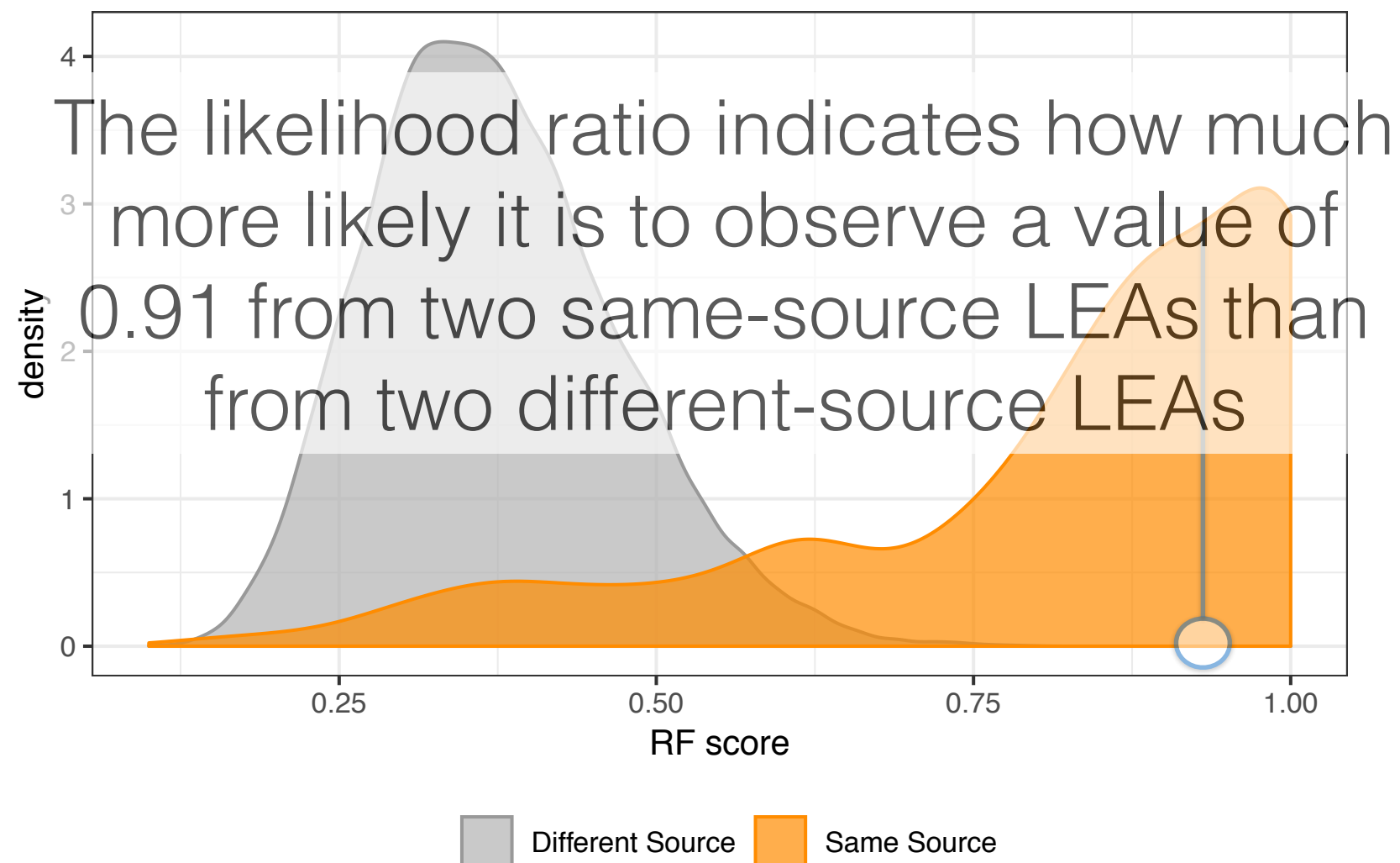


# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

○ 0.91

$$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$$

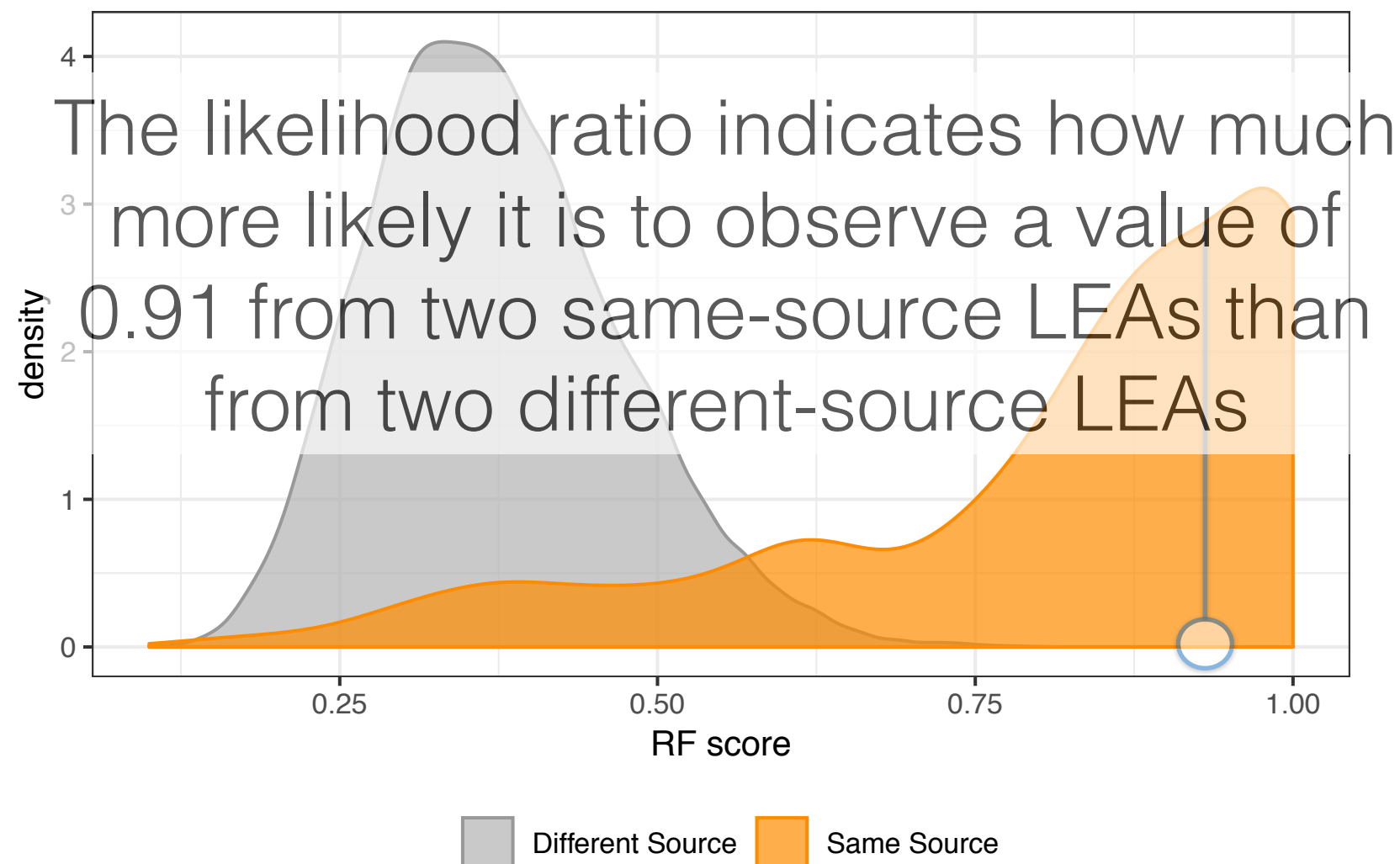


# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...

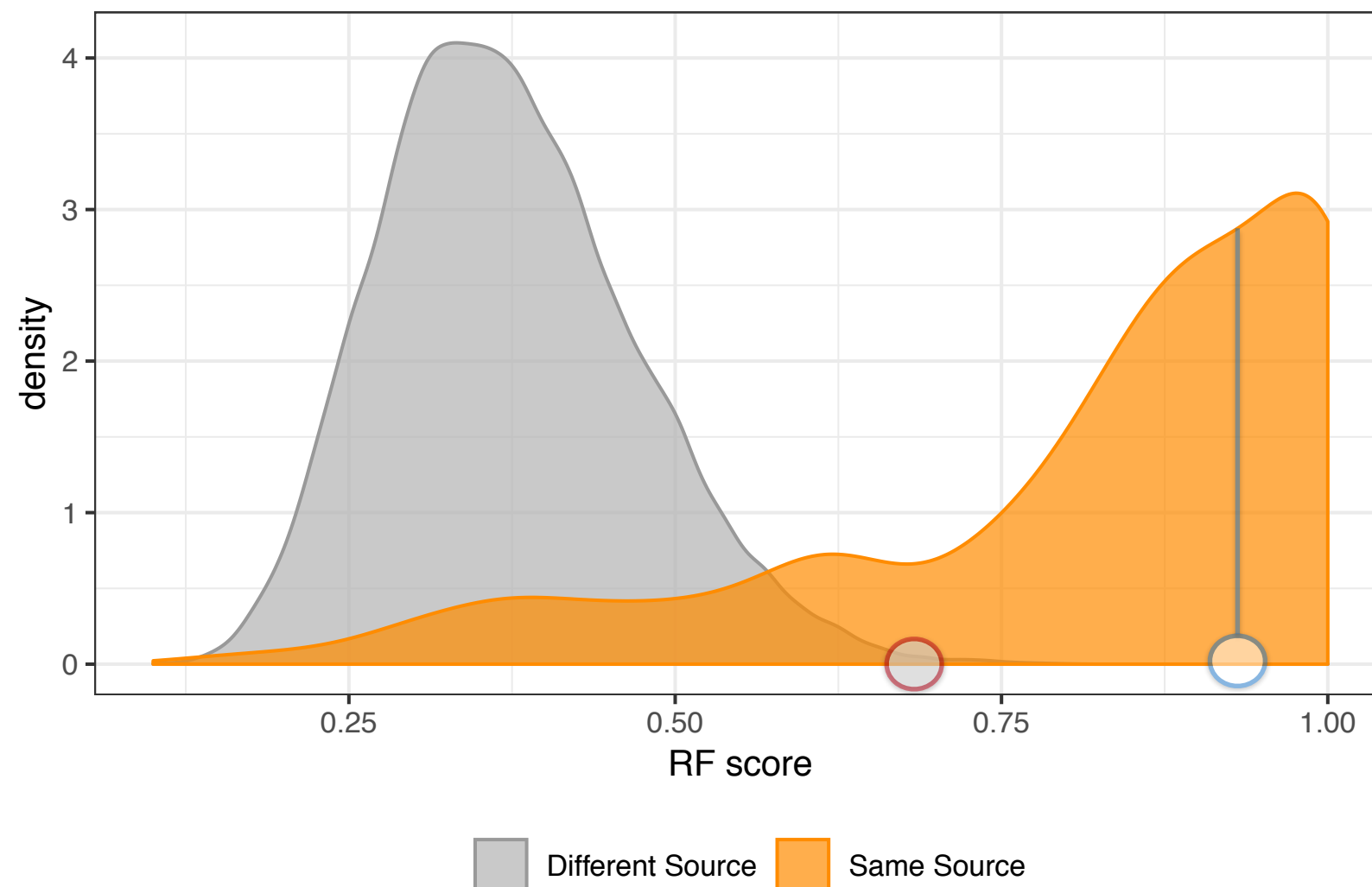
○ 0.91

$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$   
supports identification



# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...



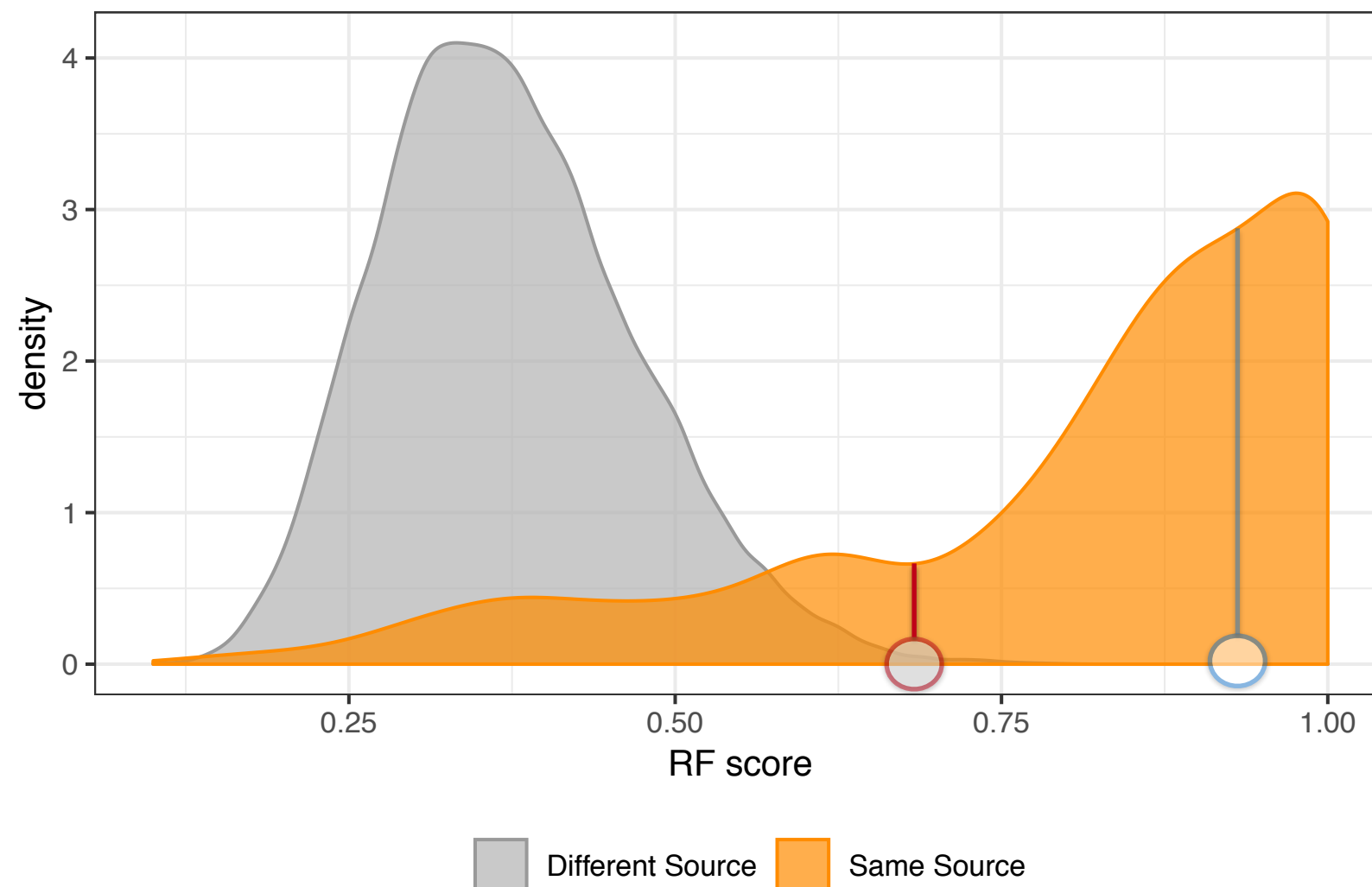
○ 0.91

$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$   
supports identification

○ 0.69

# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...



○ 0.91

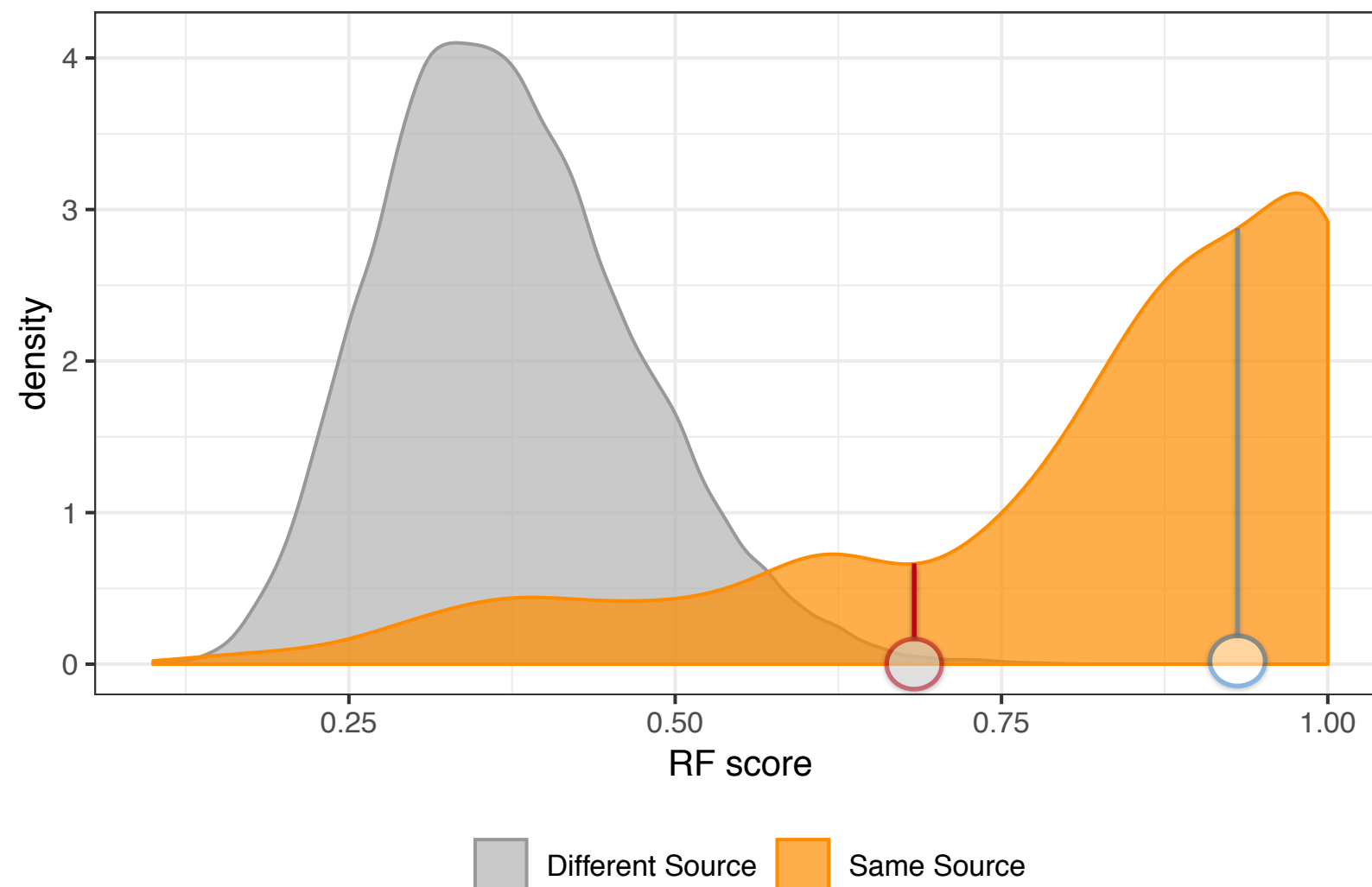
$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$   
supports identification

○ 0.69

$0.67 / 0.05 = 13.4$

# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...



○ 0.91

$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$   
supports identification

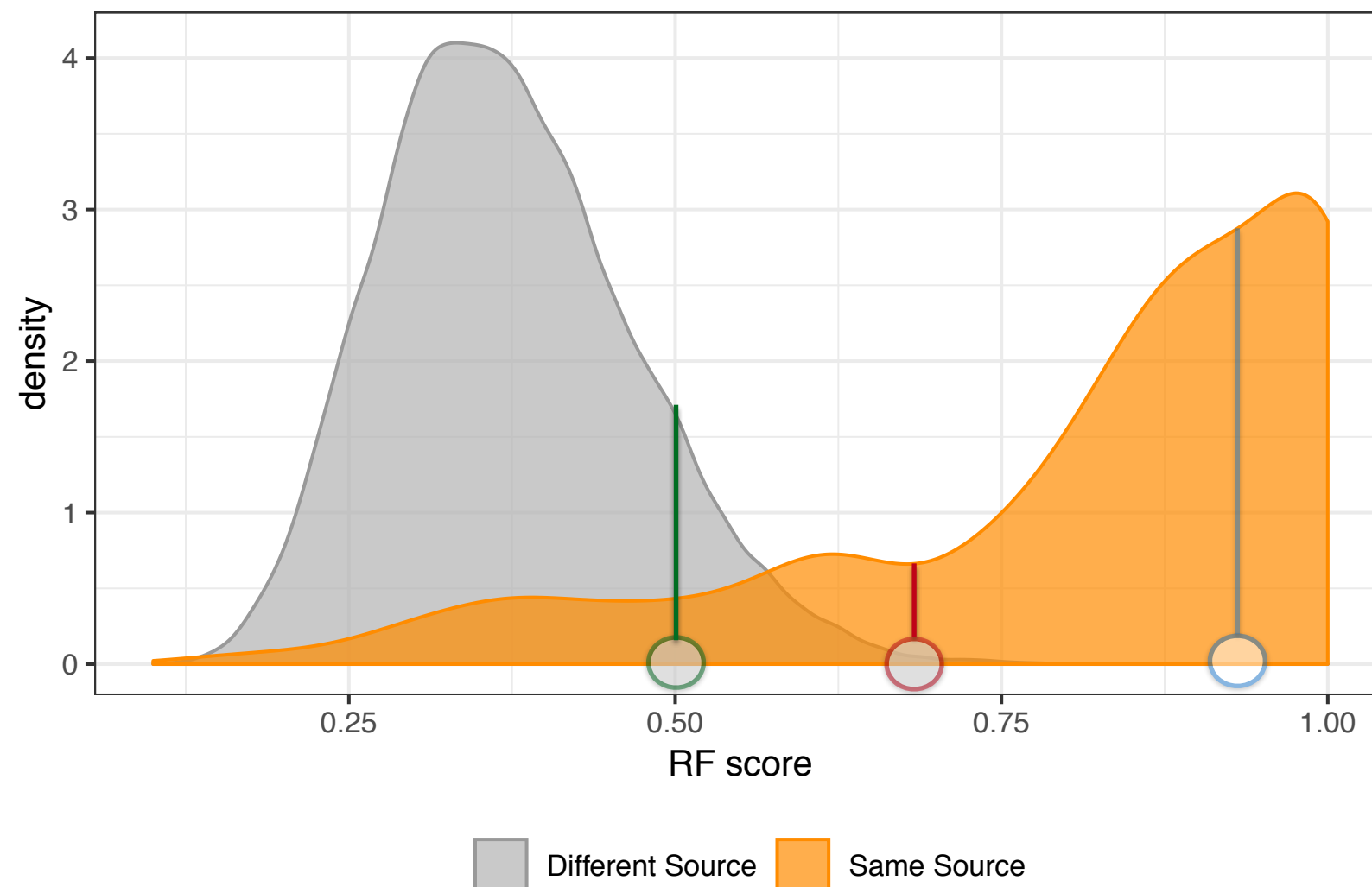
○ 0.69

$0.67 / 0.05 = 13.4$   
supports identification



# What does it mean?

★ let's assume that two bullet (land)s are compared, the model spits out a number ...



○ 0.91

$2.77 / 1.6 \times 10^{-16} = 1.7 \times 10^{16}$   
supports identification

○ 0.69

$0.67 / 0.05 = 13.4$   
supports identification

○ 0.50

$0.43 / 1.65 = 0.26$   
inconclusive

# Shades of RF Scores

# HS 224 to Clone HS 224

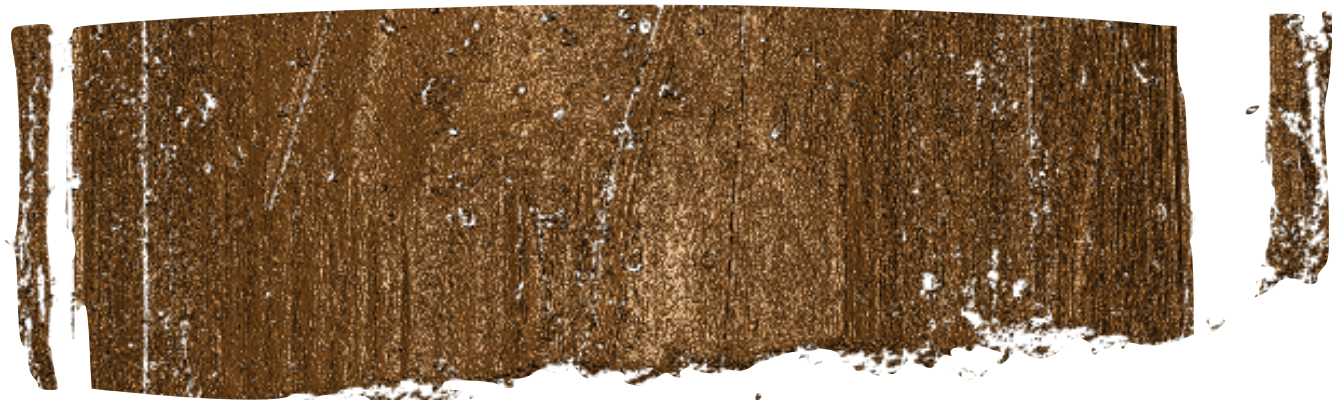
## Random Forest Scores

- ★ How well do LEAs match between a bullet and its clone?

HS 224 Br 1 - Bullet 1 - Land 1



HS 224 Clone Br 1 - Bullet 1 - Land 1





# HS 224 to Clone HS 224

## Random Forest Scores

★ How well do LEAs match between a bullet and its clone?

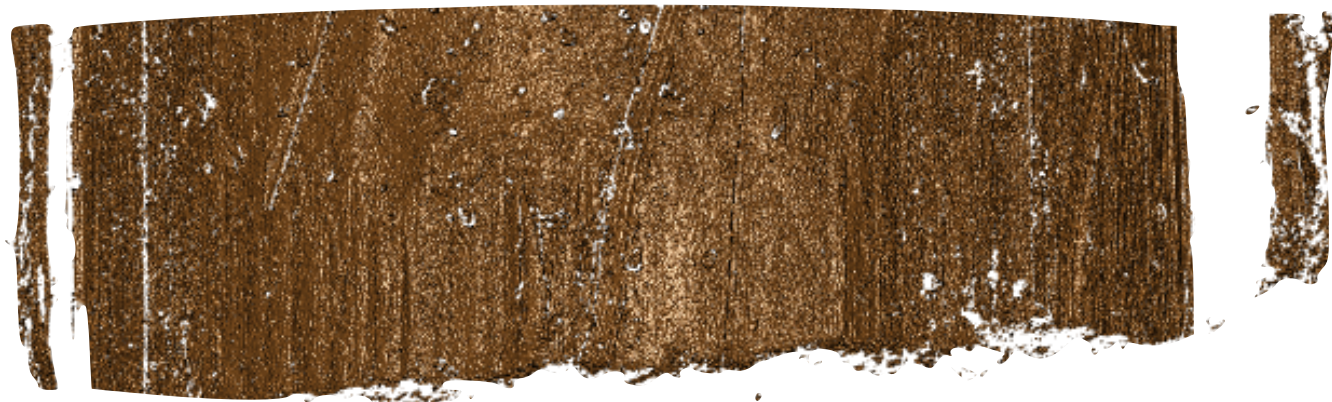
HS 224 Br 1 - Bullet 1 - Land 1



HS 224 Barrel 1 - Bullet 1 - Land 2



HS 224 Clone Br 1 - Bullet 1 - Land 1



HS 224 Clone Br 1 - Bullet 1 - Land 2





# HS 224 to Clone HS 224

## Random Forest Scores

- ★ How well do LEAs match between a bullet and its clone?

Set	Clone		Original	
	Bullet	Bullet 1	Bullet 1	Bullet 2
Clone	Bullet 1	1.00	0.80	
	Bullet 2	0.80	0.83	
Original	Bullet 1		1.00	0.98
	Bullet 2		0.98	1.00

- ★ Original bullets match with higher RF score than clones

# HS 224 to Clone HS 224

## Random Forest Scores

- ★ How well do LEAs match between a bullet and its clone?

Set	Clone		Original	
	Bullet	Bullet 1	Bullet 1	Bullet 2
Clone	Bullet 1	1.00	0.80	0.97
	Bullet 2	0.80	0.83	0.78
Original	Bullet 1		1.00	0.98
	Bullet 2		0.98	1.00

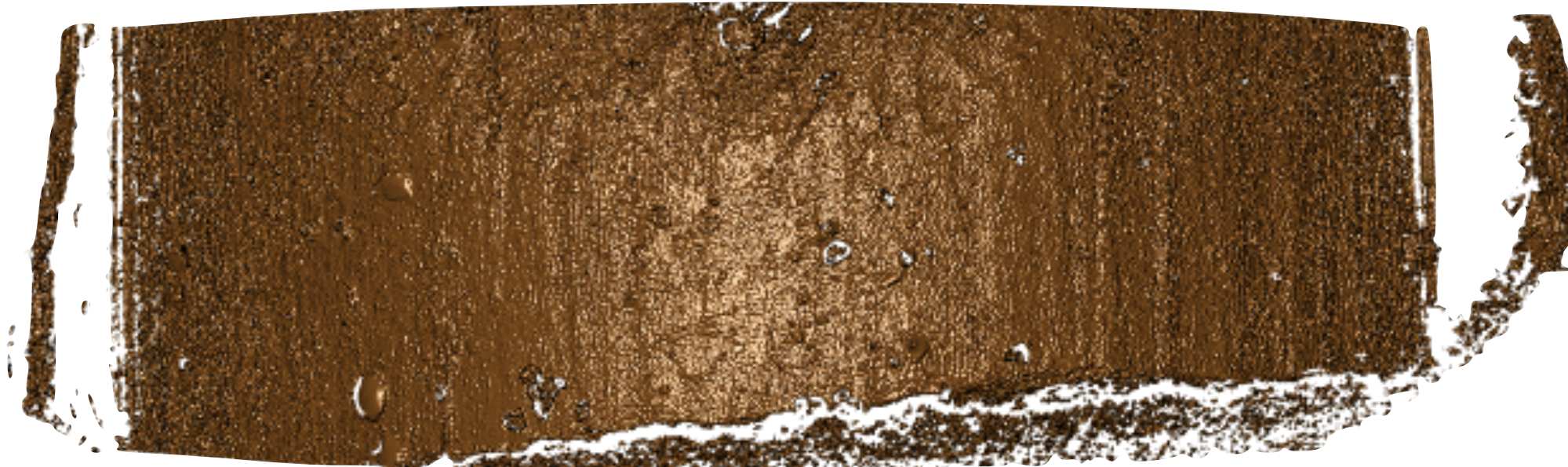
- ★ Original bullets match with higher RF score than clones
- ★ Original bullets match well to clones, almost as well as to each other



# HS 224 to Clone HS 224

## Random Forest Scores

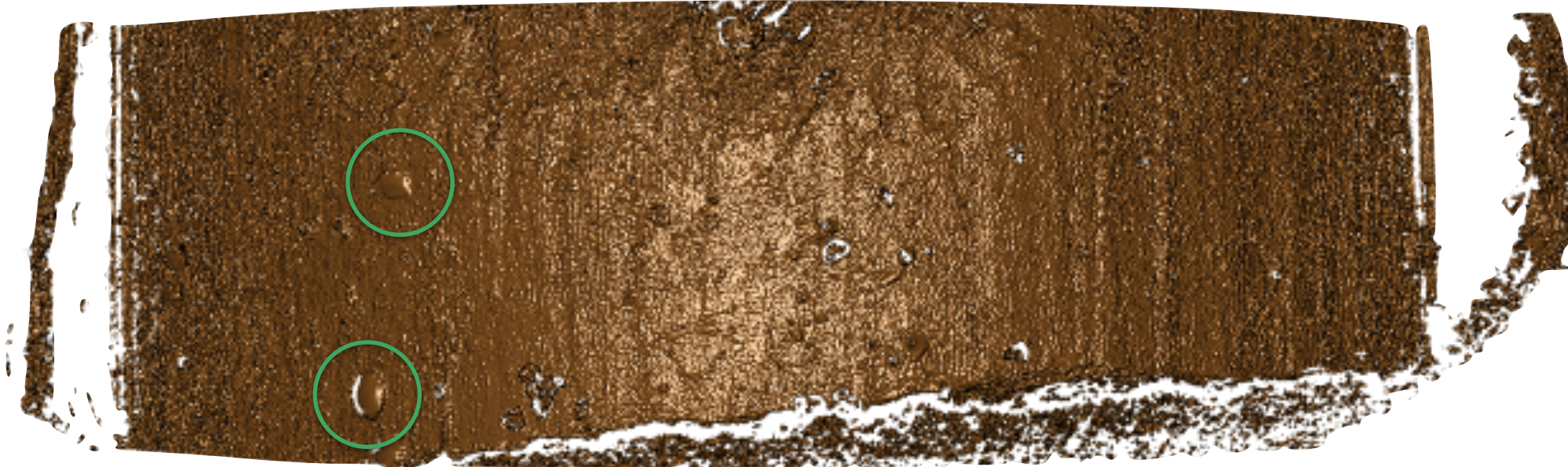
★ Some LEA scans of clones are showing artifacts:



# HS 224 to Clone HS 224

## Random Forest Scores

★ Some LEA scans of clones are showing artifacts:

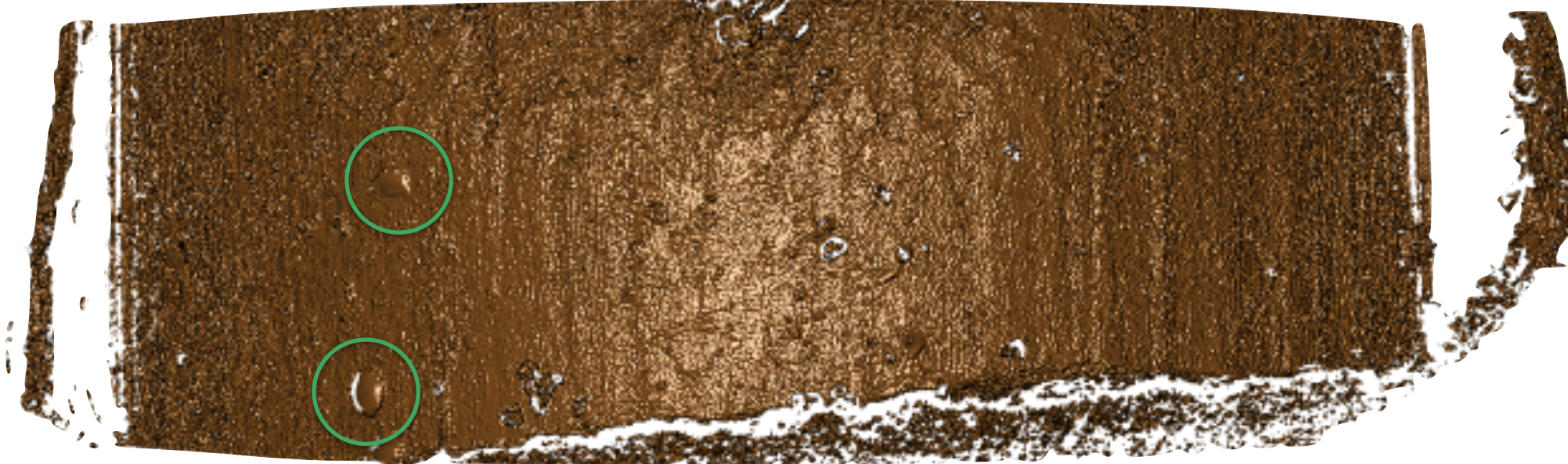




# HS 224 to Clone HS 224

## Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



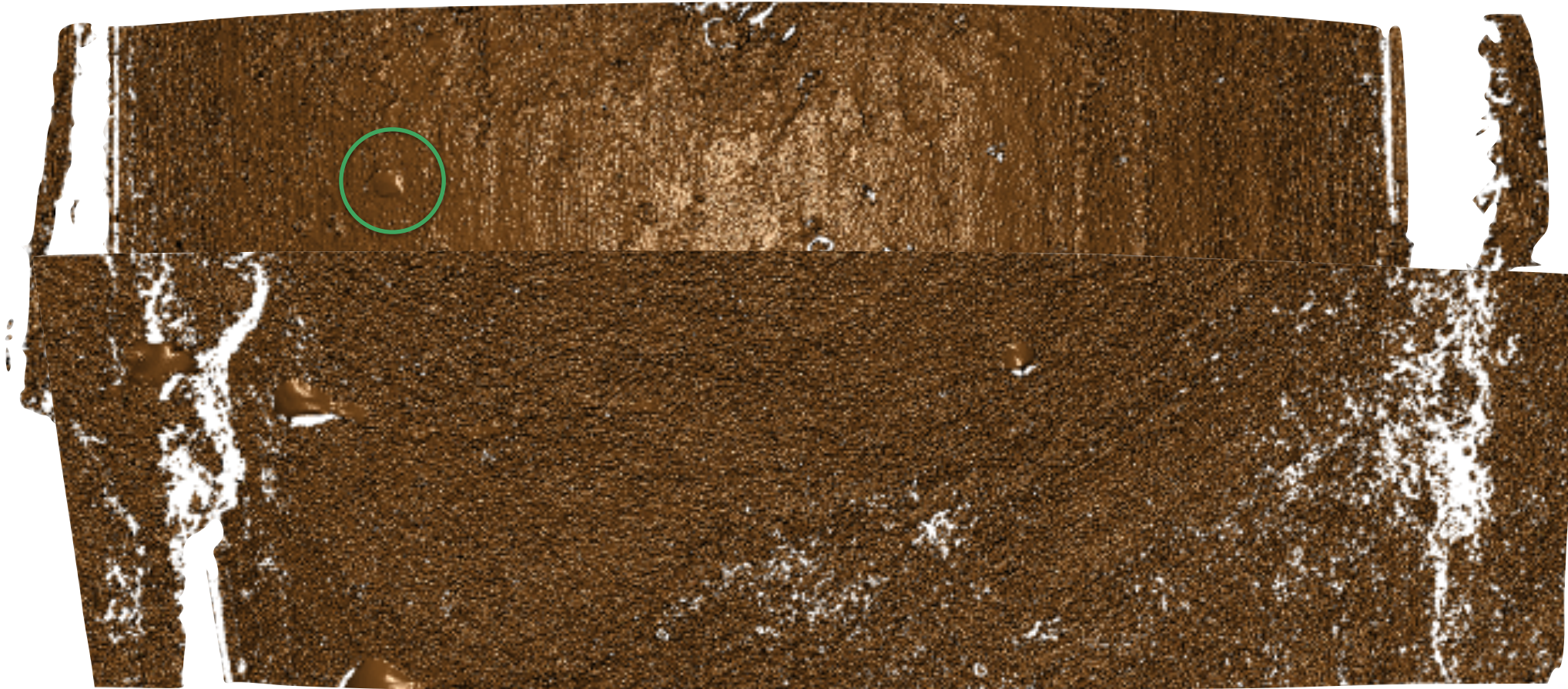
droplets



# HS 224 to Clone HS 224

## Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



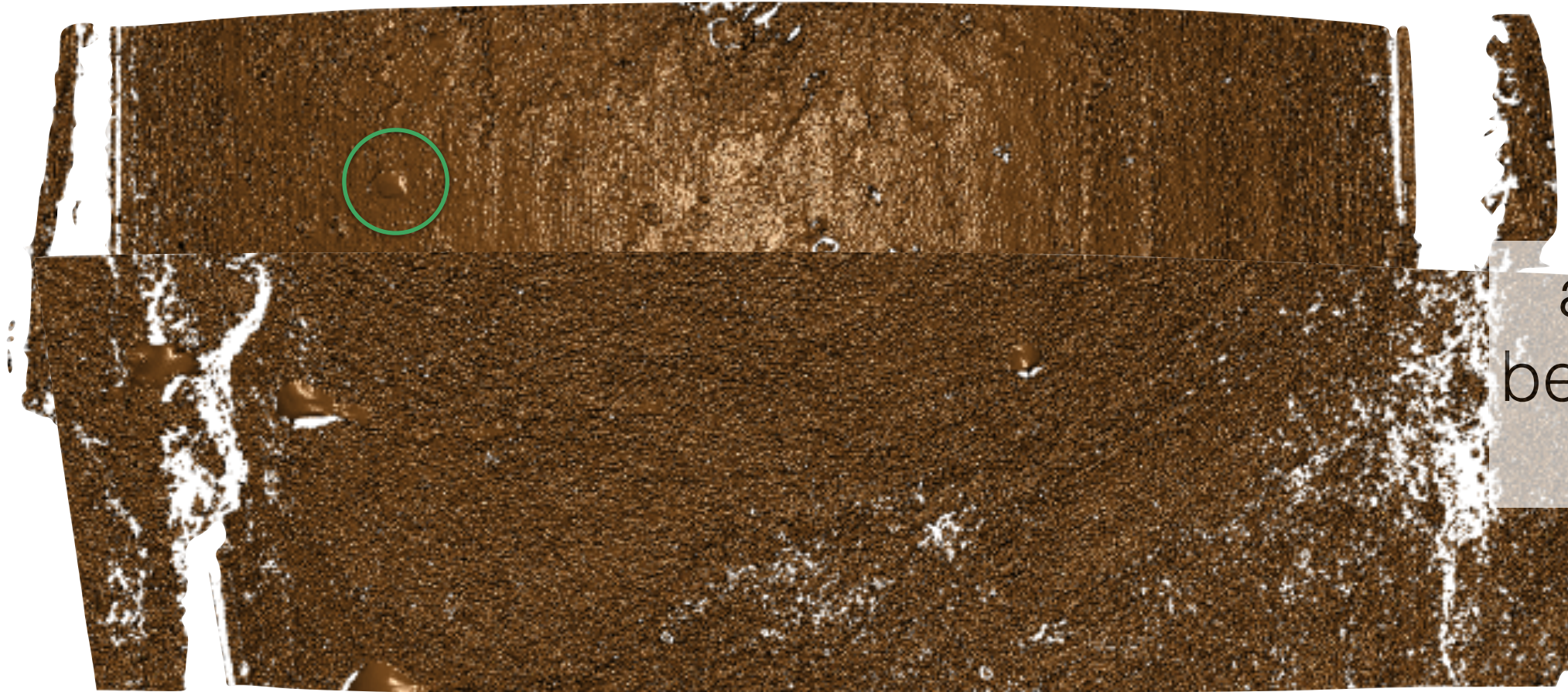
droplets



# HS 224 to Clone HS 224

## Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



droplets

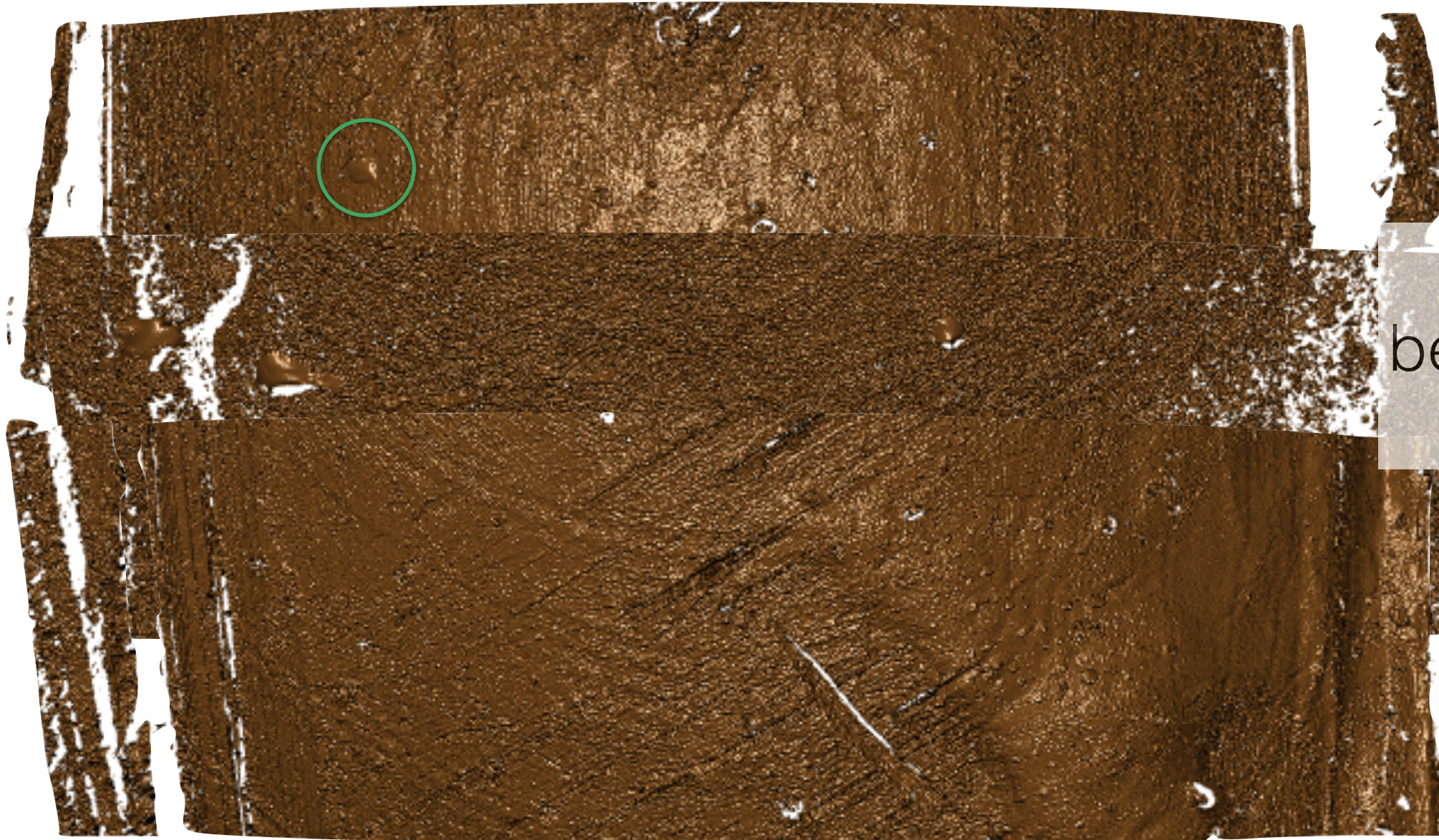
air pockets  
between clone  
and cast



# HS 224 to Clone HS 224

## Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



droplets

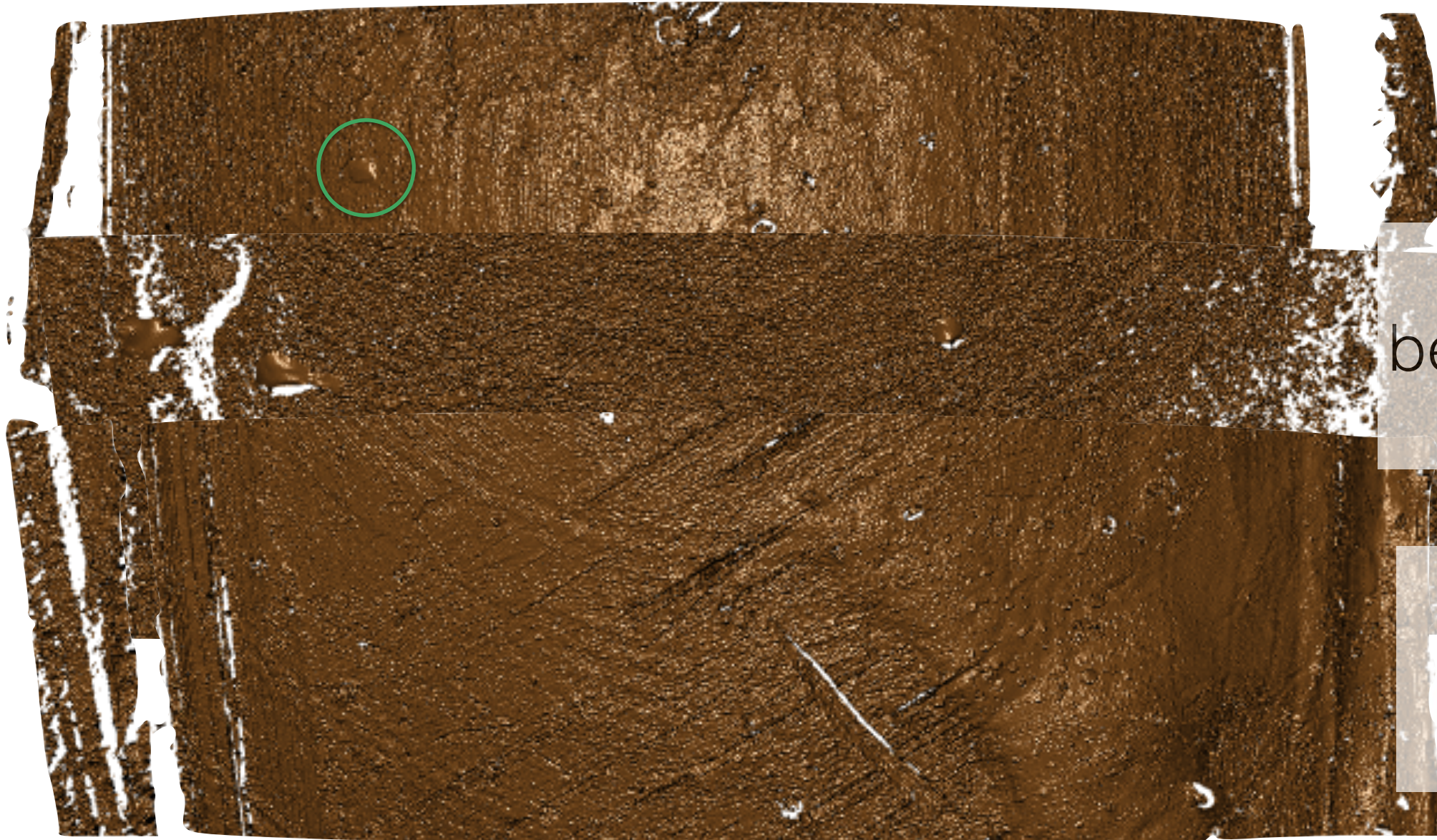
air pockets  
between clone  
and cast



# HS 224 to Clone HS 224

## Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



droplets

air pockets  
between clone  
and cast

some other  
structure but  
no striae



# HS 224 to Clone HS 224

## Random Forest Scores

★ Some LEA scans of clones are showing artifacts:



droplets

air pockets  
between clone  
and cast

Automatic RF scores allow  
us to assess clone quality  
quantitatively

some other  
structure but  
no striae

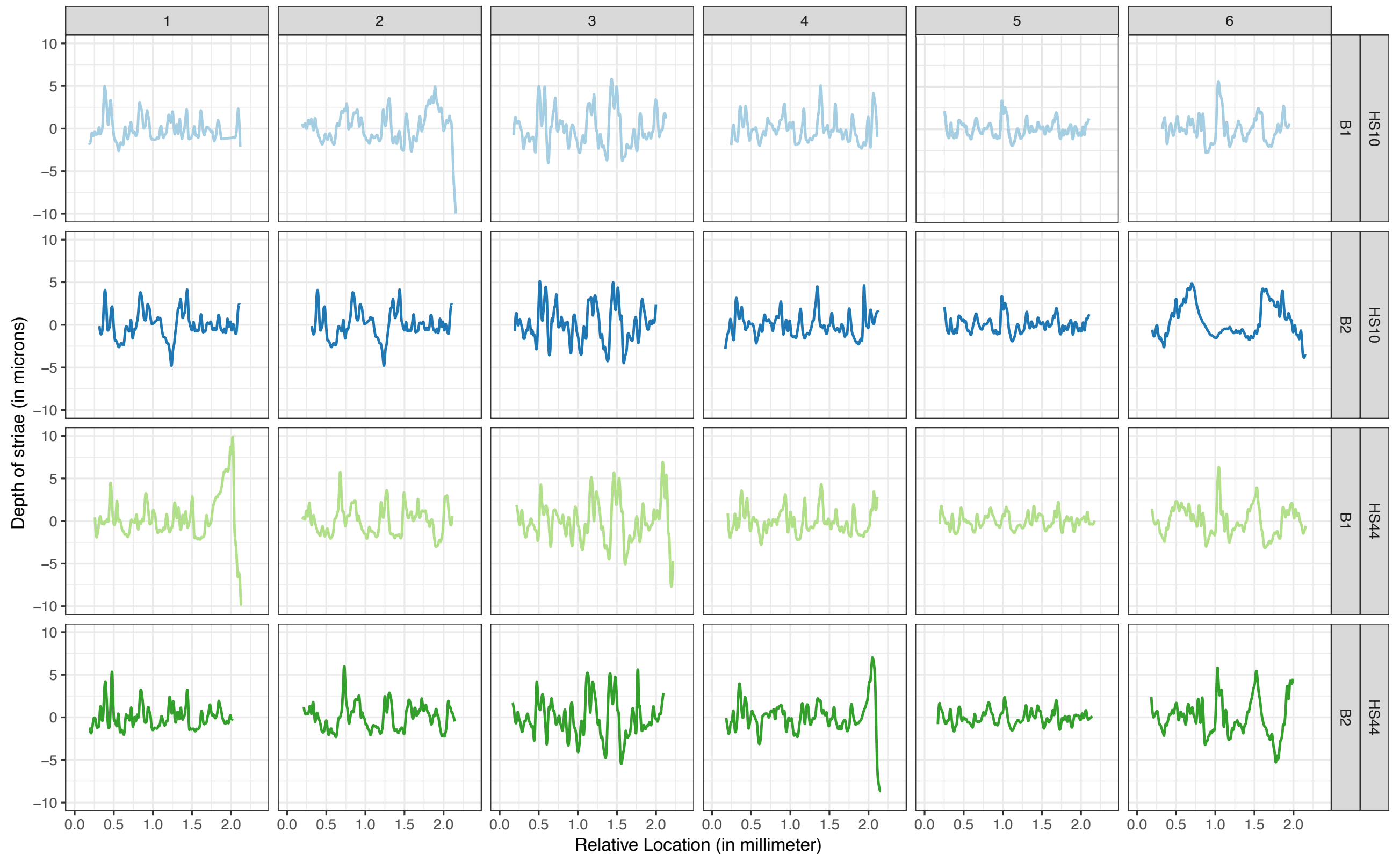


# Hamby Set 10 and 44

## Random Forest Scores

- ★ Hamby Set 10 and Set 44 are from the same 10 barrels of Ruger P-85s
- ★ Bullets from Hamby Set 44 are fired some time later (~ 240 shots in each barrel between the sets)
- ★ Two Goals:
  - ★ **check** identifications between the sets
  - ★ **quantify** identifications

# Hamby Set 10 and 44 Barrel 1



Set-Barrel-Bullet    — Br1.B1.HS10    — Br1.B2.HS10    — Br1.B1.HS44    — Br1.B2.HS44



# Hamby Set 10 and 44

## Random Forest Scores

★ Identification between Sets 10 and 44 are possible

	HS10-B1	HS10-B2	HS44-B1	HS44-B2
HS10-B1	1.00	0.65		
HS10-B2	0.65	1.00		
HS44-B1			1.00	0.79
HS44-B2			0.79	1.00

# Hamby Set 10 and 44

## Random Forest Scores

★ Identification between Sets 10 and 44 are possible

	HS10-B1	HS10-B2	HS44-B1	HS44-B2
HS10-B1	1.00	0.65	0.56	0.59
HS10-B2	0.65	1.00	0.57	0.57
HS44-B1			1.00	0.79
HS44-B2			0.79	1.00

# Hamby Set 10 and 44

## Random Forest Scores

★ Identification between Sets 10 and 44 are possible

	HS10-B1	HS10-B2	HS44-B1	HS44-B2
HS10-B1	1.00	0.65	0.56	0.59
HS10-B2	0.65	1.00	0.57	0.57
HS44-B1			1.00	0.79
HS44-B2			0.79	1.00

★ RF scores between sets are not as high as within sets

# Hamby Set 10 and 44

## Random Forest Scores

★ Identification between Sets 10 and 44 are possible

	HS10-B1	HS10-B2	HS44-B1	HS44-B2
HS10-B1	1.00	0.65	0.56	0.59
HS10-B2	0.65	1.00	0.57	0.57
HS44-B1			1.00	0.79
HS44-B2			0.79	1.00

★ RF scores between sets are not as high as within sets

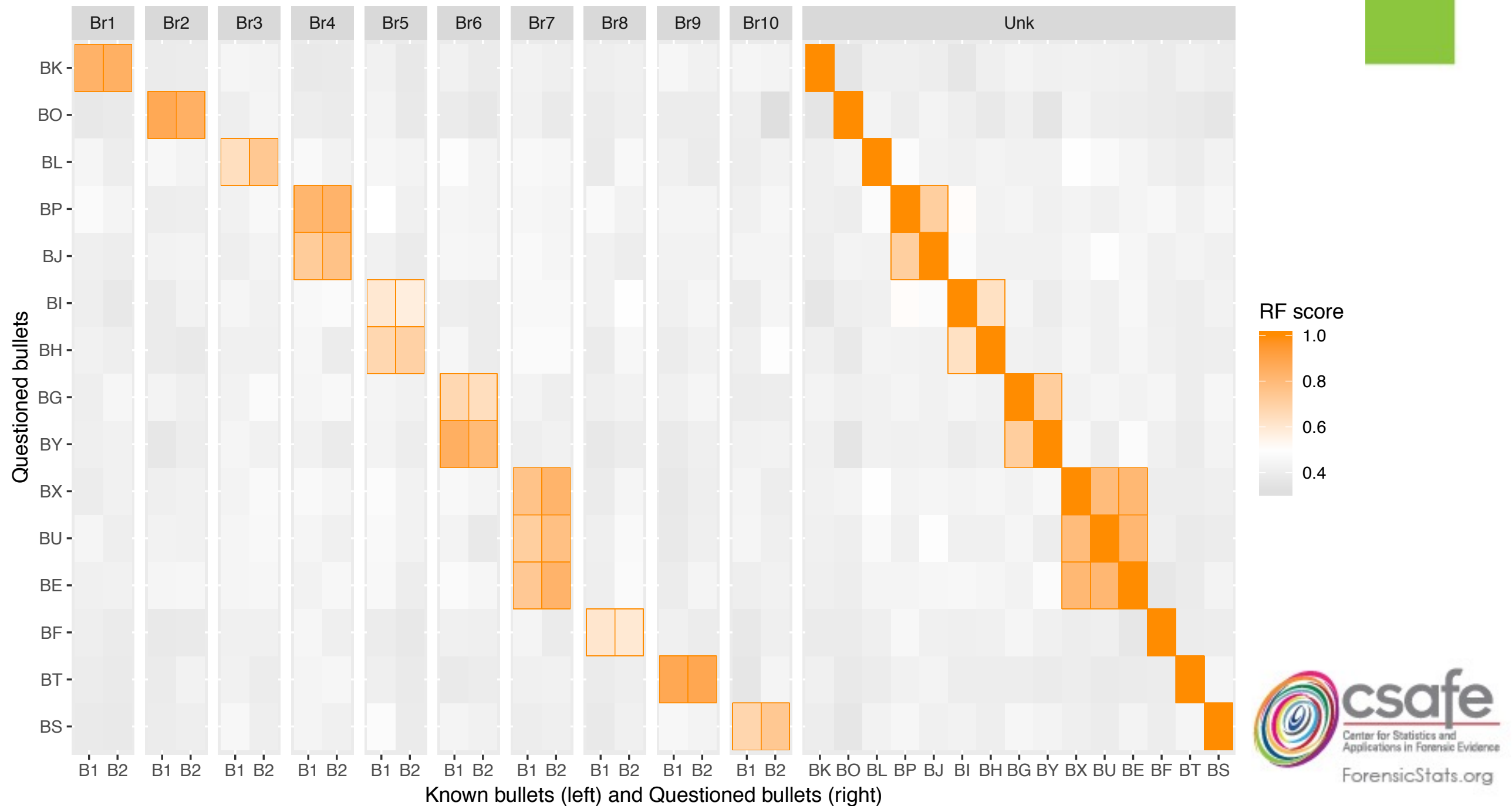
★ RF scores can be used to quantify the strength of an identification

# RF Scores in context



# Hamby Set 44

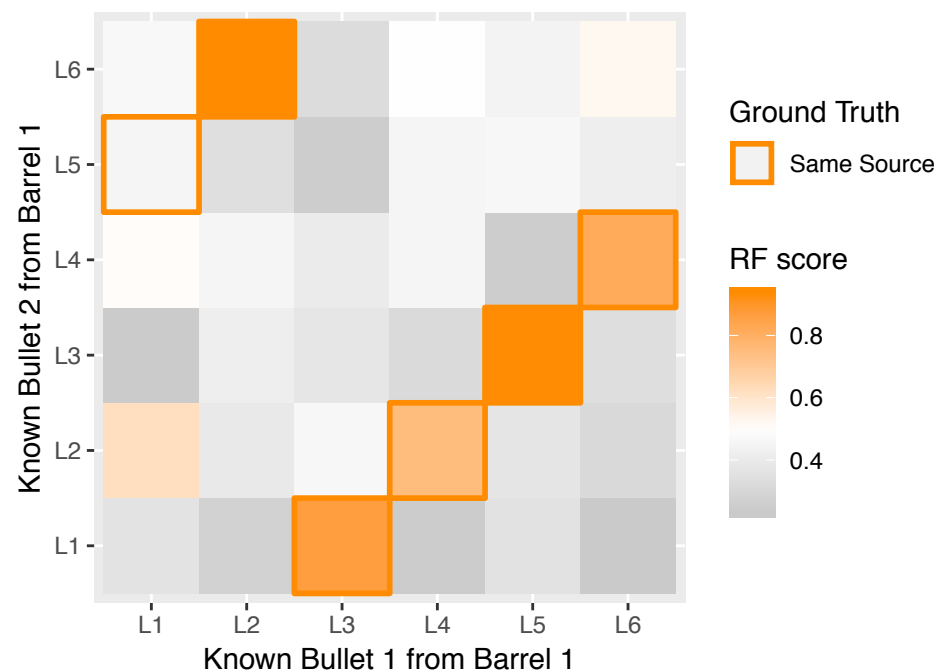
RF Scores in context of a study



# Hamby Set 44

## Random Forest Scores

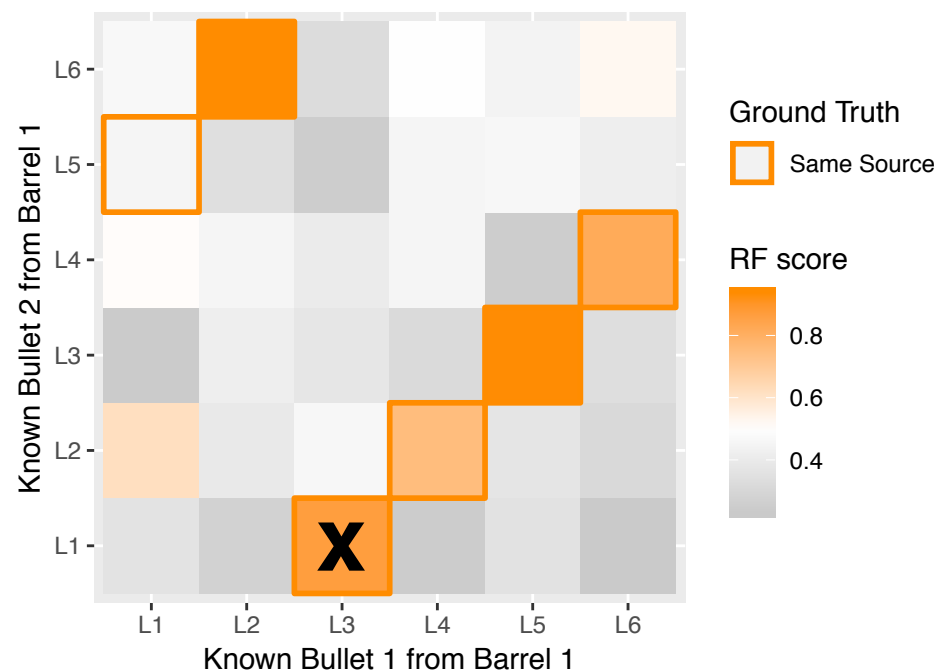
★ Land-to-land matches for bullets 1 and 2 from barrel 1



# Hamby Set 44

## Random Forest Scores

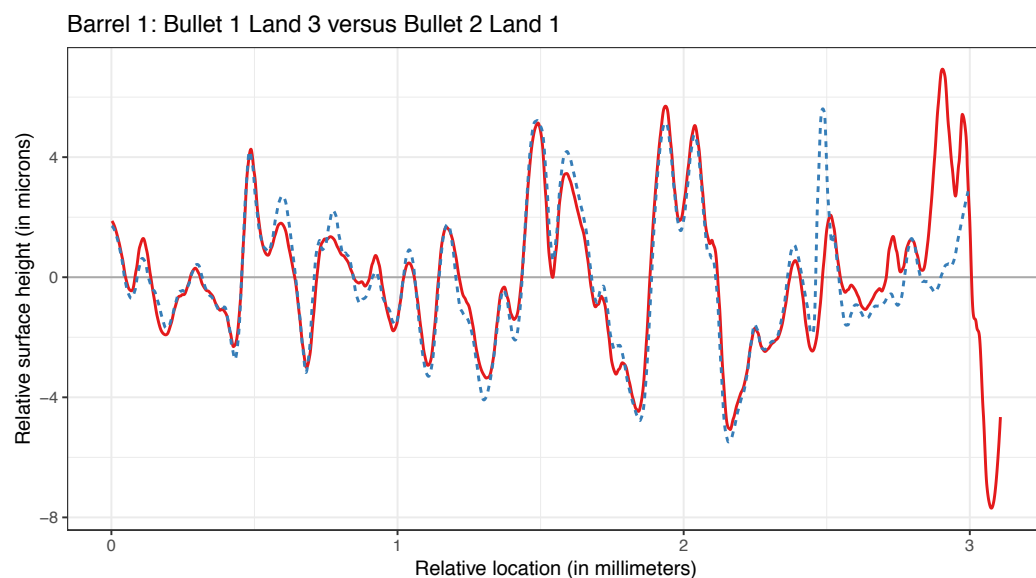
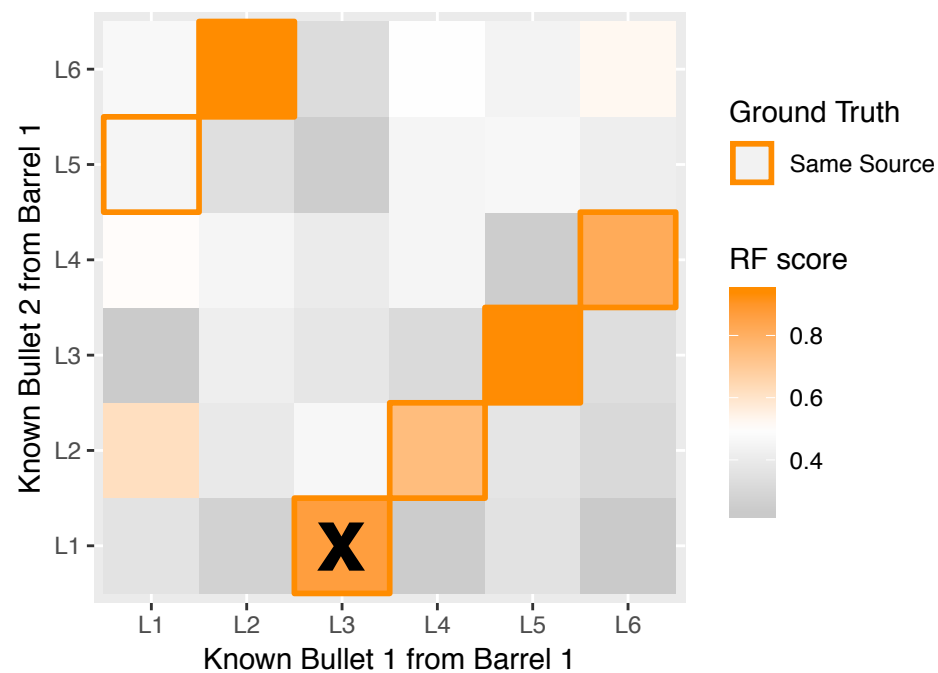
★ Land-to-land matches for bullets 1 and 2 from barrel 1



# Hamby Set 44

## Random Forest Scores

★ Land-to-land matches for bullets 1 and 2 from barrel 1

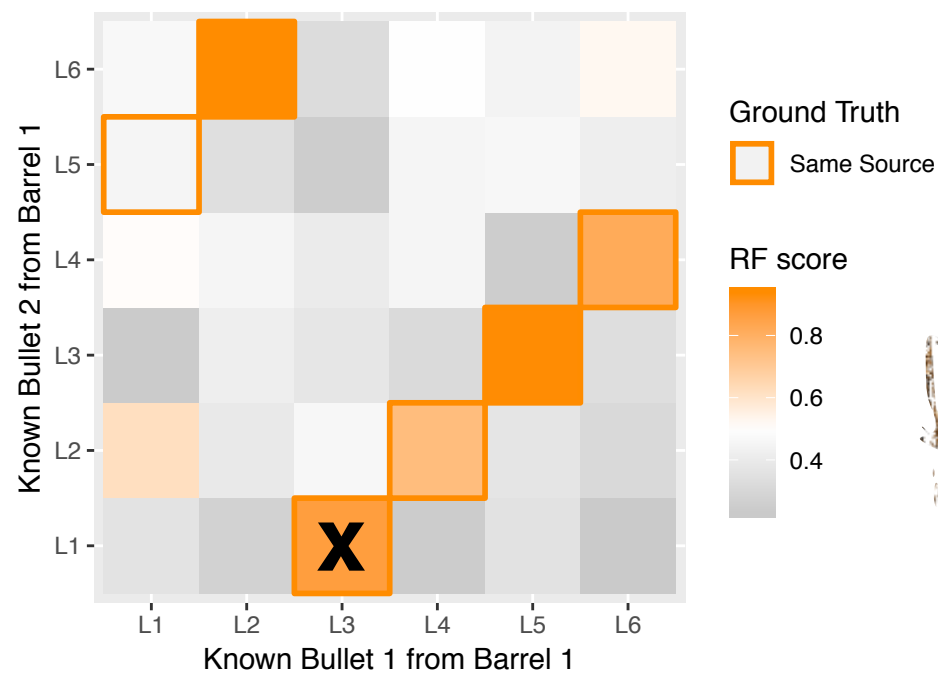




# Hamby Set 44

## Random Forest Scores

★ Land-to-land matches for bullets 1 and 2 from barrel 1



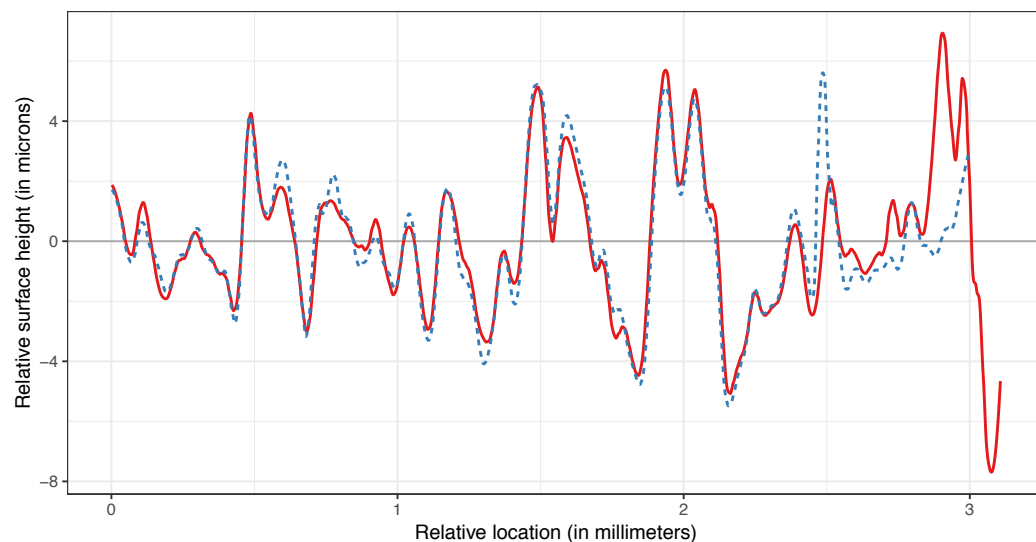
Barrel 1 - Bullet 1 - Land 3



Barrel 1 - Bullet 2 - Land 1



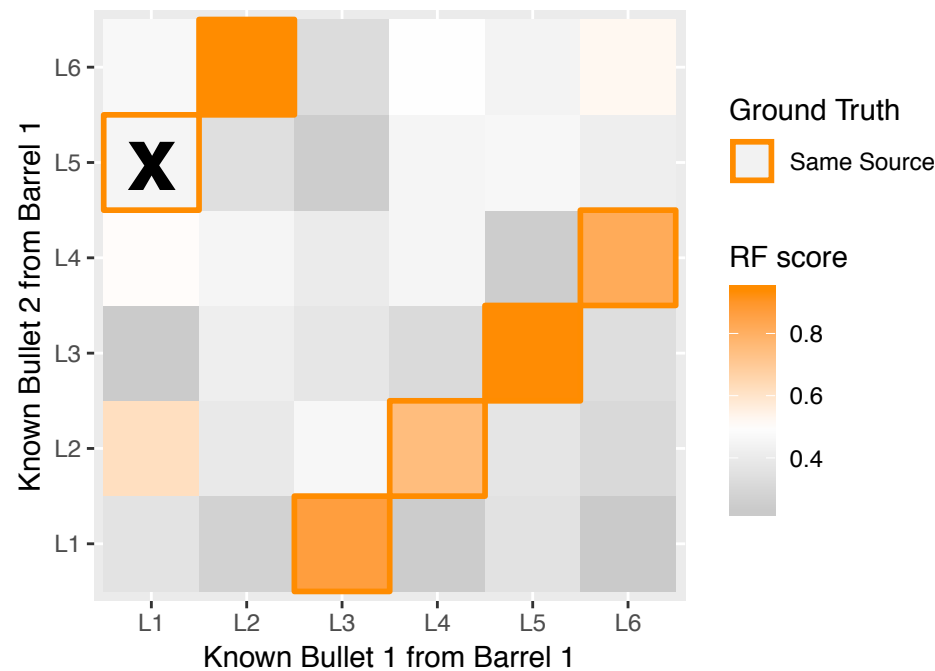
Barrel 1: Bullet 1 Land 3 versus Bullet 2 Land 1



# Hamby Set 44

## Random Forest Scores

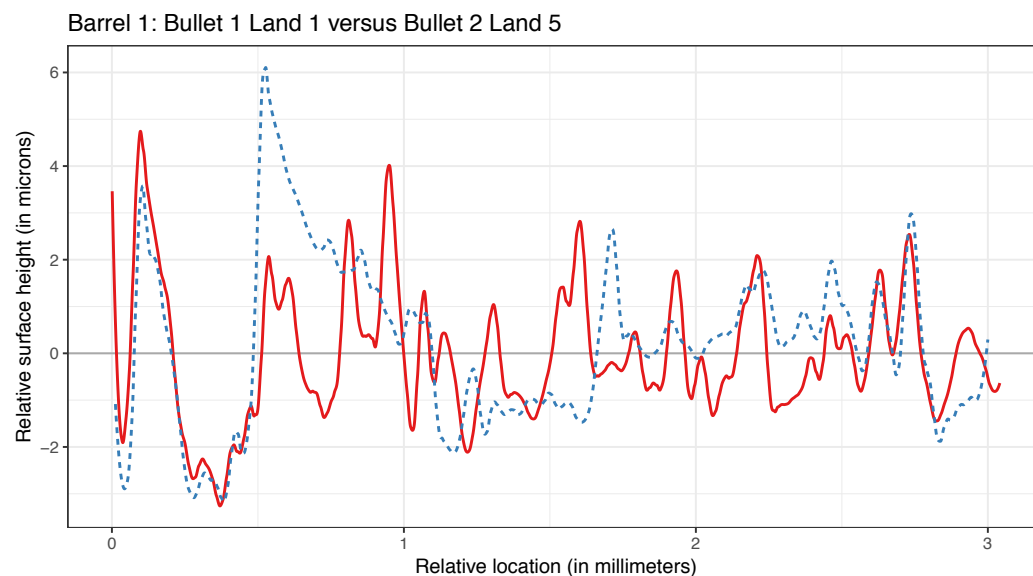
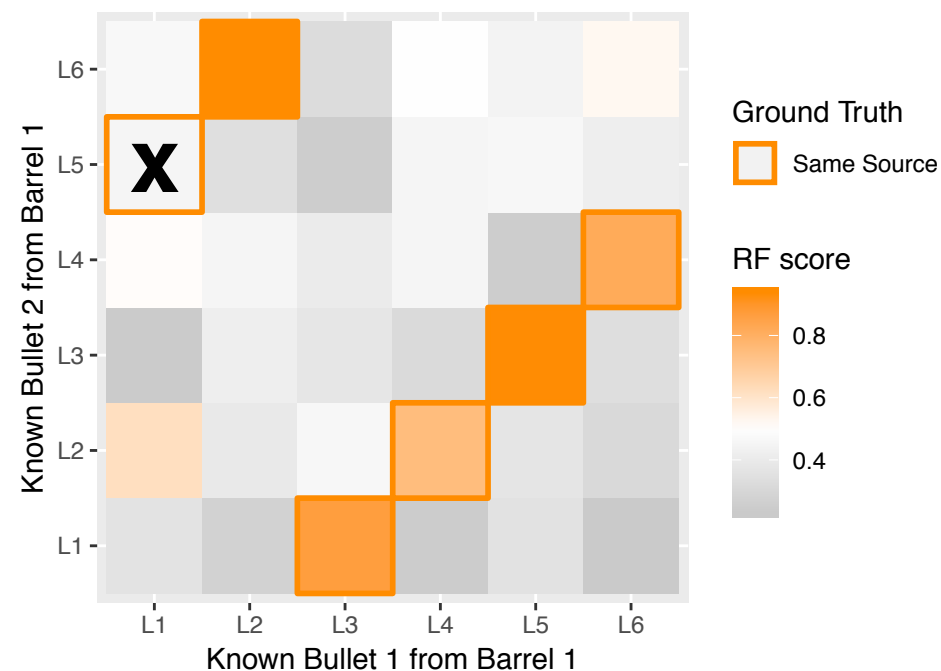
★ Land-to-land matches for bullets 1 and 2 from barrel 1



# Hamby Set 44

## Random Forest Scores

★ Land-to-land matches for bullets 1 and 2 from barrel 1

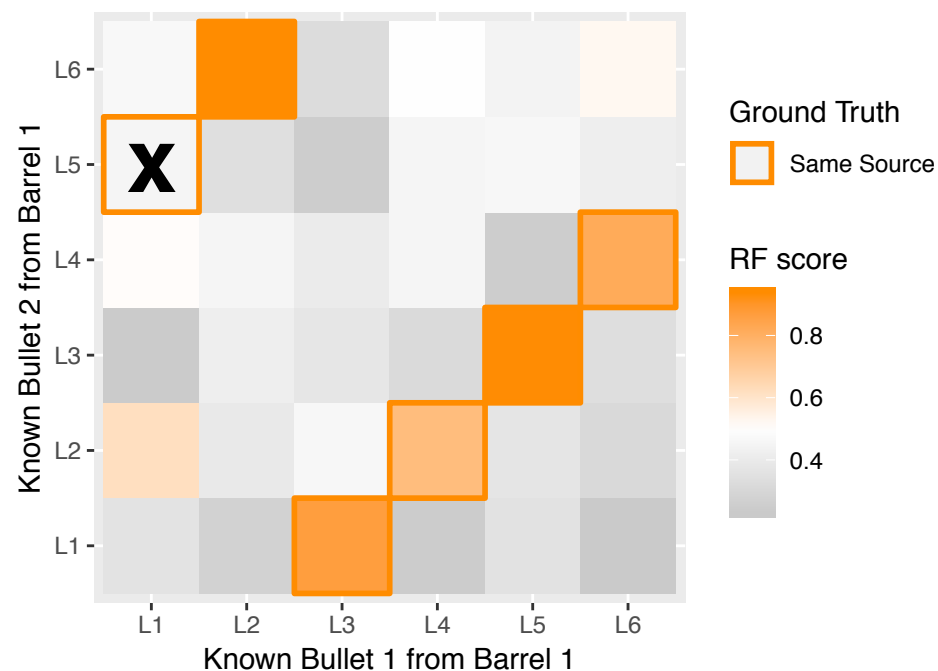




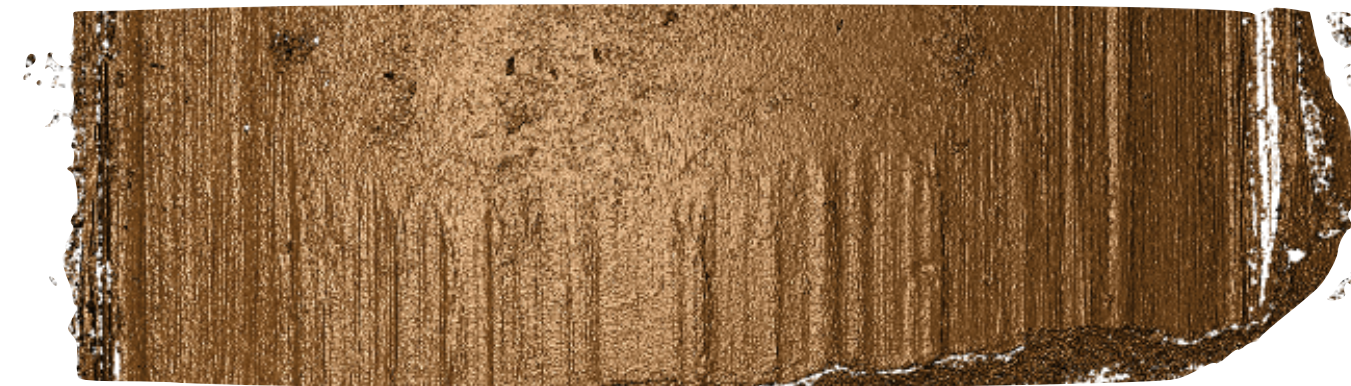
# Hamby Set 44

## Random Forest Scores

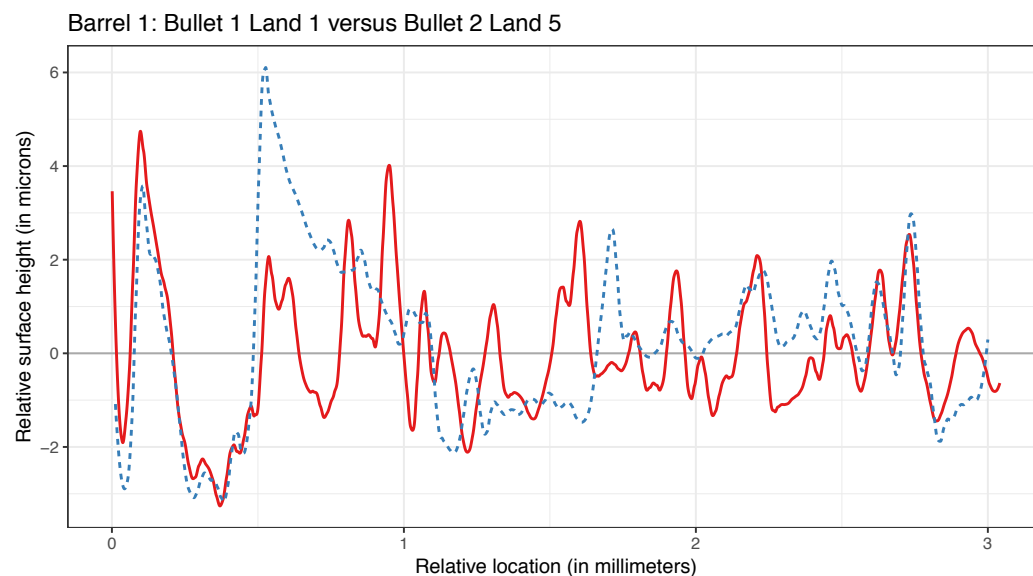
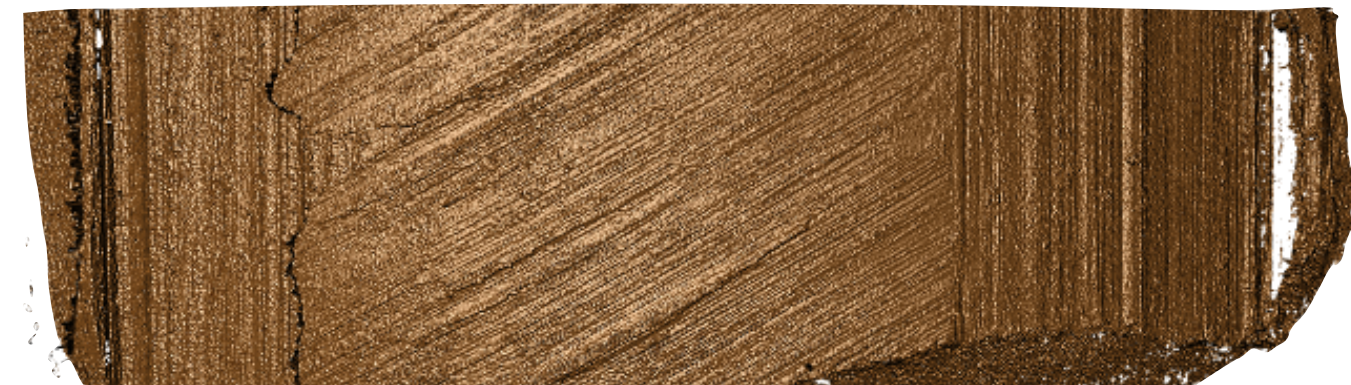
★ Land-to-land matches for bullets 1 and 2 from barrel 1



Barrel 1 - Bullet 1 - Land 1



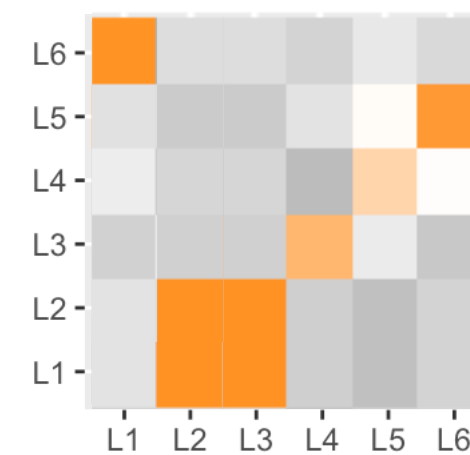
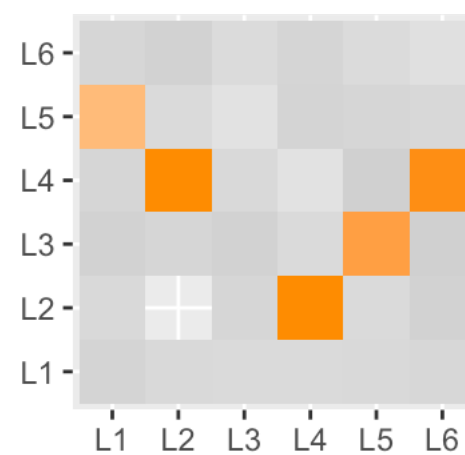
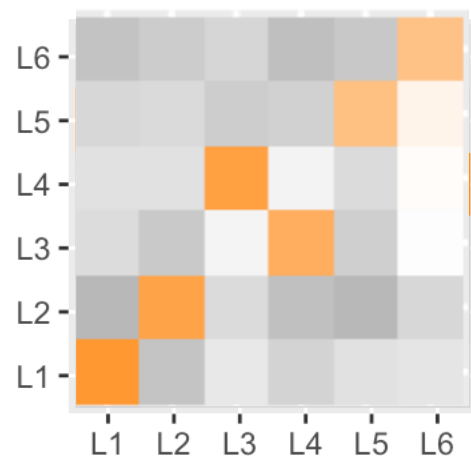
Barrel 1 - Bullet 2 - Land 5





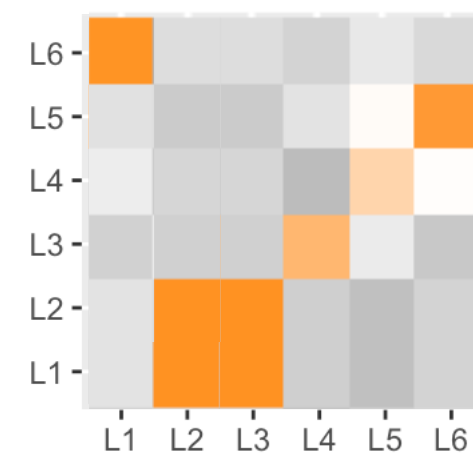
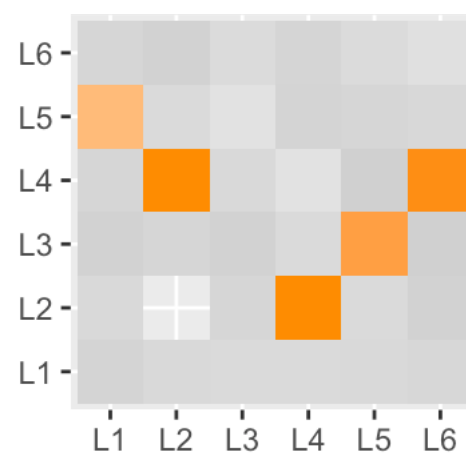
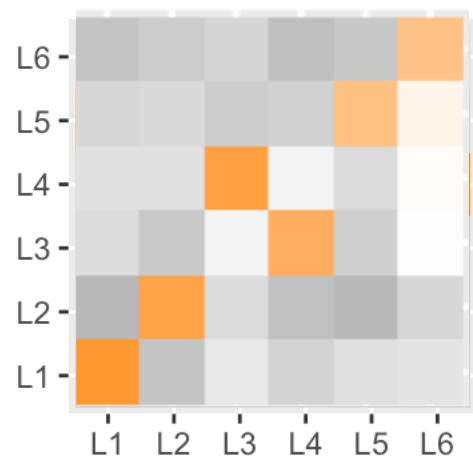
# Using the RF Score for diagnostics

★ Land-to-land matches for two bullets show oddities



# Using the RF Score for diagnostics

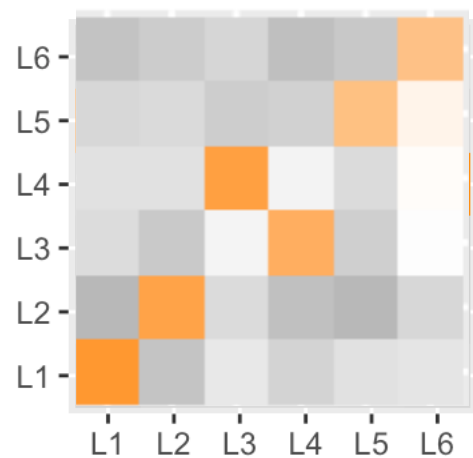
★ Land-to-land matches for two bullets show oddities



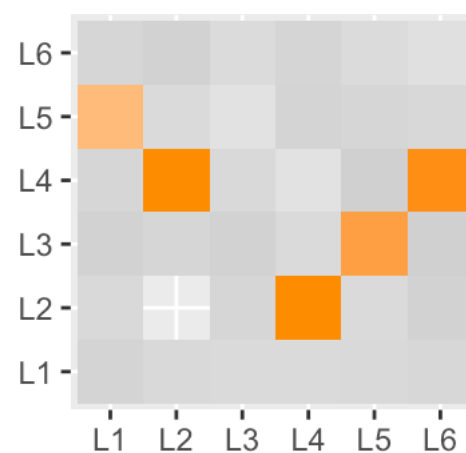
*scans for lands 3  
and 4 were mis-  
labelled for one  
bullet*

# Using the RF Score for diagnostics

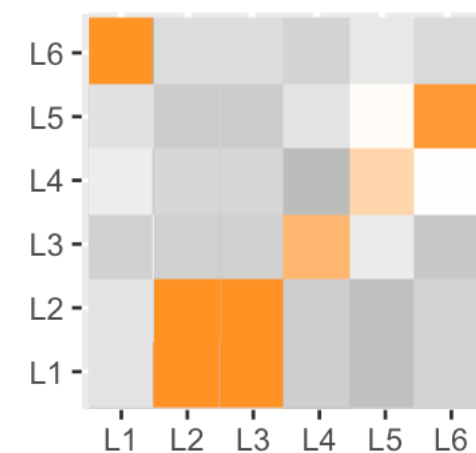
★ Land-to-land matches for two bullets show oddities



*scans for lands 3  
and 4 were mis-  
labelled for one  
bullet*



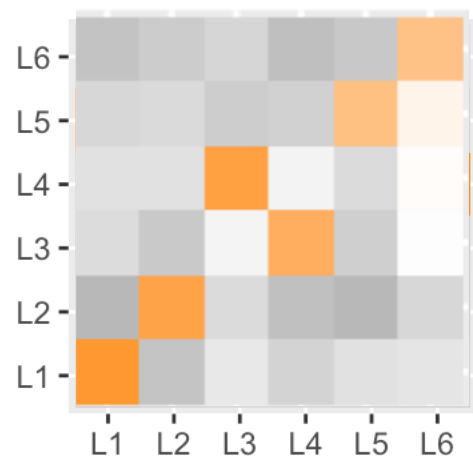
*instead of turning  
right to scan land 6,  
scanner turned left  
and scanned land 4  
again*



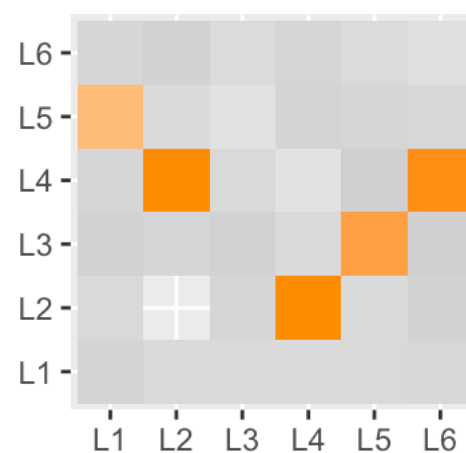


# Using the RF Score for diagnostics

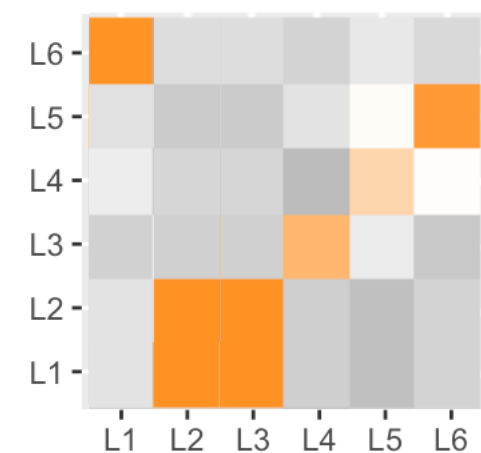
★ Land-to-land matches for two bullets show oddities



*scans for lands 3 and 4 were mislabelled for one bullet*



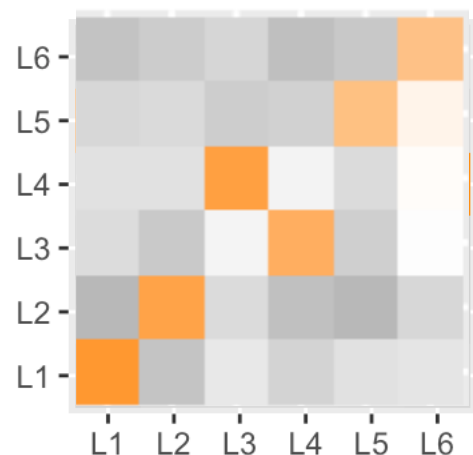
*instead of turning right to scan land 6, scanner turned left and scanned land 4 again*



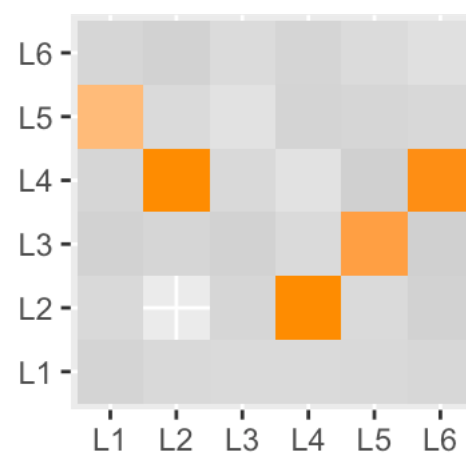
*scans for land 2 and land 3 are identical*

# Using the RF Score for diagnostics

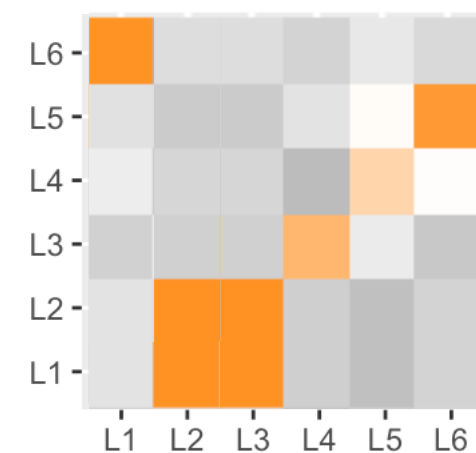
★ Land-to-land matches for two bullets show oddities



*scans for lands 3 and 4 were mislabelled for one bullet*



*instead of turning right to scan land 6, scanner turned left and scanned land 4 again*



*scans for land 2 and land 3 are identical*

These things went wrong in our scanning lab

# Summary & further work

- ★ Random Forest Model provides interpretable and consistent scores for quantifying identifications
  - ★ matching clones to original bullets - assess clone quality
  - ★ matching bullets between Hamby sets
  - ★ quantification of bullet and scan deficiencies
- ★ Visualizations with increased context used for diagnostics
- ★ Limitations: traditional rifling only (LEAs are matched)

# Thank You!

## Questions?

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ISU CSAFE bullet team



<https://forensicstats.org/>

@csafe\_coe 